CTRLSUM: Towards Generic Controllable Text Summarization

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Abstract

Current summarization systems yield generic summaries that are disconnected from users' preferences and expectations. To address this limitation, we present CTRLSUM, a generic framework to control generated summaries through a set of keywords. During training keywords are extracted automatically with-007 out requiring additional human annotations. At test time CTRLSUM features a control function to map control signal to keywords; through engineering the control function, the 011 same trained model is able to be applied 012 013 to control summaries on various dimensions, while neither affecting the model training process nor the pretrained models. We additionally explore the combination of keywords and text prompts for more control tasks. Ex-017 periments demonstrate the effectiveness of CTRLSUM on three domains of summariza-019 tion datasets and five control tasks: (1) entitycentric and (2) length-controllable summarization, (3) contribution summarization on scientific papers, (4) invention purpose summarization on patent filings, and (5) question-guided summarization on news articles. Moreover, when used in a standard, unconstrained summarization setting, CTRLSUM is comparable 027 or better than strong pretrained systems.

1 Introduction

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Neural summarization systems aim to compress a document into a short paragraph or sentence while preserving key information. There are two common categories of summarization systems: extractive summarization where models find and copy important portions of a documents (Cheng and Lapata, 2016; Nallapati et al., 2017; Narayan et al., 2018), and abstractive summarization where models freely generates novel sentences (Rush et al., 2015; See et al., 2017; Paulus et al., 2018). The focus of this work is on abstractive summarization.

In a common setting, summarization methods take a source document as input and yield a generic

summary which covers content selected arbitrarily by the model. However, to be useful, automatically generated summaries should cover content considered important by the readers. For example, Figure 1 shows an NBA basketball news article, and the reference summary aggregates the results of several matches. Yet fans of certain players or teams might only be interested in the matches and statistics which include their entities of interest. 043

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Motivated by this observation, in this work we focus on controllable summarization which allows the users to control the generated summaries. We propose CTRLSUM, a framework to control summaries through a set of keywords. At training time, the model learns to predict summaries conditioned on both the source document and keywords, which are easily identified from training summaries. During inference, a *control function* is designed depending on the specific control aspect to map user preference to keywords to control the summary.

While simple, keywords provide a clean separation of test-time user control and the training process - different dimensions of the generated summary could be controlled through engineering the test-time control function, while the training process and pretrained model remain unchanged. In contrast, typical controllable summarization methods (Fan et al., 2018; Liu et al., 2018; Tan et al., 2020; Suhara et al., 2020) predefine the specific control aspect (e.g. entity, length, topic, etc.) and rely on the corresponding control annotations during training. Consequently, they require to train a separate model for each control aspect and cannot generalize to new control aspects at test time. Figure 1 demonstrates such distinction. The training process in CTRLSUM is agnostic to the specific control aspect, leading to possibly generic controllable summarization with a single model.

CTRLSUM is closely related to recent keywordguided summarization methods. While they mainly focus on improving the summary quality in tra-



Figure 1: **Top Right:** traditional methods which incorporate the specific control aspect of user interest into training process, and thus requires training a separate model for each aspect. **Bottom Right:** the proposed CTRLSUM framework, where the model training relies on automatic keywords and is separated from the control aspect. At test time a specially designed control function maps control signal to keywords, and a single trained model achieves controllable summarization on different dimensions.

ditional, unconstrained summarization tasks (Li et al., 2018; Elsahar et al., 2020; Saito et al., 2020b; Dou et al., 2021), or only study a specific control aspect like length control (Saito et al., 2020a), we generalize keyword-guided summarization as a generic controllable summarization framework, and explore its novel applications to a wide range of control tasks in this paper. In addition, inspired by the multi-task ability of language models through prompting (Radford et al., 2019; Brown et al., 2020), we further study the possible combination of keywords and prompts in CTRLSUM for more generic control purposes.

We use pretrained BART (Lewis et al., 2019) as the underlying architecture and perform experiments on three datasets: CNN/Dailymail news articles (Hermann et al., 2015), arXiv scientific papers (Cohan et al., 2018), and BIGPATENT patent documents (Sharma et al., 2019). We quantitatively evaluate CTRLSUM on five control aspects: (1) entity-centric ($\S4.2$) and (2) length-controllable summarization (§4.3), (3) summarizing the contributions of scientific papers, (4) summarizing the purpose of an invention (§4.4), and (5) summarizing answers to given questions in a zero-shot reading comprehension setting ($\S4.5$). Notably, our approach also achieves comparable or superior performance to the strong BART baseline on all datasets in a standard, unconstrained setting (§4.6).

We release our trained CTRLSUM checkpoints and demos¹ to encourage others to try CTRL-SUM for their own control of interest. On a higher level, we expect the exploration in this paper to attract attentions to the controllable summarization task, and provide a springboard for future research on personalized summarization systems. 116

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2 CTRLSUM

2.1 Formulation

Unconstrained neural summarization models are trained to learn the conditional distribution $p(\mathbf{y}|\mathbf{x})$, where \mathbf{x} and \mathbf{y} represent the source document and summary respectively. In this setting, generated summaries depend solely on the document \mathbf{x} without requiring human involvement. To control the content of generated summaries, we propose using additional keywords \mathbf{z} to represent user preferences and training a model that predicts the conditional distribution $p(\mathbf{y}|\mathbf{x}, \mathbf{z})$. As shown in Figure 1, keywords act as an interface between users and an otherwise black-box neural model.

Formally, at training time keywords are identified using source documents and ground-truth summaries while at test time they are a function of source documents and user control signal:

$$\mathbf{z}_{\text{train}} = f(\mathbf{x}, \mathbf{y}), \qquad \mathbf{z}_{\text{test}} = g_{\text{control}}(\mathbf{x}, \mathbf{c}), \quad (1)$$

where f denotes the automatic keyword extraction algorithm, **c** is the user control signal, and $g_{control}$ represents the control function. **c** can be instantiated as different forms of control signals (e.g., entity words or length value) as users or control tasks need during test stage while not requiring re-training the model. The control function admits different designs depending on the control task. We detail $g_{control}$ next, but describe the automatic keyword extraction later in §2.3.

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¹Code, data, and the trained model checkpoints are provided in the supplementary material. Demo will be released after the review period.

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2.2 The Control Function g_{control}

 $g_{\text{control}}(\mathbf{x}, \mathbf{c})$ maps the user control signal **c** to spe-150 cific keywords. Given a control task (e.g. entity or length control), g_{control} can be designed specif-152 ically to perform the task, for example, in entity 153 control $\mathbf{c} = e$ is the entity of interest and g_{control} 154 may directly output e as the keyword; in length control $\mathbf{c} = l$ is the desired summary length value, in 156 this case g_{control} may automatically extract a certain number (which depends on l) of keywords from the 158 source document. This way, g_{control} provides clean 159 separation of test-time user control and the train-160 ing process – through only engineering the control function, the same trained model can be adapted to 162 new control tasks without changing model parameters or the training process. Such a framework is 164 generic and broadly applicable to many control as-165 pects, which stands in contrast to previous work on 166 controllable summarization (Fan et al., 2018; Liu et al., 2018) which pre-define one control aspect and design a specific training procedure for it. In \$3 we will show how the same model is applied to 170 five different controllable summarization tasks.

A note on the scope of focus: Mathematically, $g_{\text{control}}(\mathbf{x}, \mathbf{c})$ covers a use case where the user may directly input arbitrary keyword sequences at test time.² However, such an open-ended setting is poorly defined as a control task - the specific control aspect is unconstrained and unclear, which brings challenges to the evaluation that is already difficult in controllable generation due to lack of references. Therefore, we focus on several specific, well-defined control tasks as to be shown in §3. In this paper we do not explore CTRLSUM as an open-ended system, but leave it as future work.

Automatic Keyword Extraction 2.3

As shown in Eq. 1, the training keywords z_{train} depend on both the source x and reference y, while \mathbf{z}_{test} are manipulated according to the control signal. We further provide $g_{\text{control}}(\mathbf{x}, \mathbf{c})$ with an *optional* automatic keyword extraction method at inference time, which allows CTRLSUM to perform unconstrained summarization when the control signal $\mathbf{c} = \emptyset$. It may also be useful in control tasks where automatic keywords from x need to be first extracted. Below we describe them respectively:

At Training Time: We first greedily select sentences from the document that maximize the

ROUGE scores (Lin, 2004) with the reference summary. This step constrains keywords to those found in important sentences. Next, we identify all of the longest sub-sequences in the extracted sentences that have matched sub-sequences in the groundtruth summary. Finally, we remove duplicate words and stop words and keep the remaining tokens as keywords. Compared to typical keywords extraction methods (Riloff and Lehnert, 1994; Mihalcea and Tarau, 2004) which output only a few salient words, our extraction retains most content words found in the summary. This is to build a reliable correlation between their presence in the input and the target, ensuring that keywords are not ignored by the model at test time.

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At Inference Time: We formulate keyword extraction as a sequence labeling task. Concretely, we train a BERT-based sequence tagger (Devlin et al., 2018) on the keywords and documents from training dataset. This tagger then computes the selection probability q_i for each token in the test document. Similar to training time extraction, we first select n_s sentences with the highest average token selection probability. Within these sentences words with $q_j > \epsilon$ are selected as keywords up to a maximum number of m_{max} . $n_s, \epsilon, m_{\text{max}}$ are hyperparameters selected based on the unconstrained summarization performance on validation datasets, and the results are reasonably robust to different values, as shown in Appendix G.

Training Details 2.4

We train the CTRLSUM model to maximize $p(\mathbf{y}|\mathbf{x}, \mathbf{z})$. Specifically, we prepend the keyword sequence z to the source document x separated with a special token =>, then the resulted sequence is taken as the input to the model to predict the summary. Our training procedure is model-agnostic so that CTRLSUM can employ any sequence-tosequence architecture as the backbone without modifications. We note that this is much simpler than recent keyword-based summarization systems (Li et al., 2018; Saito et al., 2020a; Dou et al., 2021) where they tweak the specific model architecture to incorporate the keywords signal.

Keyword Dropout: As in §2.3, our training keyword extraction retains most content words from the summary. However, we found that such keywords are empirically too informative so that the model learns to ignore the source document while generating the summary. To remedy this, we ran-

²In this case \mathbf{c} is any sequence directly from the user and $g_{\text{control}}(\mathbf{x}, \mathbf{c}) = \mathbf{c}.$

Table 1: Qualitative examples from the output of CTRLSUM. Left column shows source or the generic reference summary. Keywords are bolded. "[]" denote that the tokens are used as both keywords and prompts. The special symbol "|" in keywords represents the sentence boundary of the keywords' positions in the source document, see Appendix A.1 for details on this.

Source or Reference	Control Aspect	Keywords (bolded) or Prompts and Model Output
Source: Hundreds of additional Iraqi troops are being sent to reinforce colleagues who are trying to fend off ISIS' attempt to overrun Iraq's largest oil refinery, a key paramilitary force		ISIS – The reinforcements come four days after ISIS began attacking Baiji oil refinery.
said Tuesday. The reinforcements come four days after ISIS began attacking northern Iraq's Baiji oil refinery, a key strategic resource that has long been a target because the facility refines much of the fuel used by Iraqis domestically. The additional	Entity	Hasd Al-Shaabi – The reinforcements come from Camp Speicher, a fortified Iraqi base near Tikrit. They include two federal police regiments, an Iraqi military quick reaction force battalion and a regiment from Hasd Al-Shaabi.
troops came from Camp Speicher, a fortified Iraqi base near the city of Tikrit, according to the media office of the Hasd Al- Shaabi militia. The reinforcements include two federal police regiments, an Iraqi military quick reaction force battalion and a regiment from Hasd Al-Shaabi. [ignoring 110 tokens] The refinery is 40 kilometers (25 miles) from Tikrit.	Length	(Length bucket 0) Iraqi troops ISIS oil refinery come days attacking Baiji refinery base Tikrit – The reinforcements come four days after ISIS began attacking Iraq's Baiji oil refin- ery. The additional troops came from Camp Speicher, a forti- fied Iraqi base near Tikrit.
remely is to knoheers (25 miles) from Tikit.	QA	[Q: Where did the additional troops come from? A:] – Camp Speicher near the city of Tikrit.
<i>Reference:</i> multiple sclerosis is a degenerative condition of the central nervous system that affects nearly 2.5 million of individuals in terms of their [ignoring 60 tokens]. in this work we resort to regularized machine learning methods for binary classification and multiple output regression we propose a pipeline that can be used to predict the disease progression from patient reported measures . the obtained model is tested on a data set collected from an ongoing clinical research project .	Contribution	[the main contributions of this paper are : (1)] – we propose a regularization - based data analysis pipeline for the prediction of the transition point between relapsing - remitting and secondary - progressive forms of multiple sclerosis ; (2) we test the performance of the proposed pipeline on a data set collected from patients currently enrolled in an ongoing funded project ; (3) we discuss the potential role of patient reported outcome measures (_ proms _) in the study of the disease evolution .
<i>Reference:</i> an implantable intraocular pressure sensor system has a sealed geometric shape with an internal pressure at a first value. the sealed geometric shape has a first light permitting surface and a second flexible surface. a pair of photocells is located in the sealed geometric shape. a light shield is coupled to the second flexible surface . when the second flexible sur- face is deflected, a light measurement by the pair of photocells indicates an intraocular pressure condition.	Purpose	[the purpose of the present invention is] – to provide an in- traocular pressure measuring system that is small enough to be implanted in the eye of a patient

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domly drop keywords at training time so that the model learns to rely on keywords, while also learning to still carry over key information from the source document. Note that keyword dropout is applied at training time only.

3 The Example Control Tasks

In this section we define five controllable summarization tasks that we study in this paper and the respective control function g_{control} . Qualitative examples of each task are shown in Table 1.

3.1 Through Keywords

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Entity Control: The goal is to produce summaries that focus on specified entities of interest. We directly use the entity word as keyword, i.e. $g_{\text{control}}(\mathbf{x}, \mathbf{c} = \mathbf{e}) = \mathbf{e}$ where \mathbf{e} is the entity.

Length Control: We allow manipulation of the summary length through a user-specified length parameter. Specifically, we first separate the training 264 data into 5 buckets by summary length so that each bucket has the same number of examples. Then we compute the average number of keywords K_l for each bucket on the training data. At test time, a user 268 can specify a length value $\mathbf{c} = l \in \{0, 1, 2, 3, 4\}$ and g_{control} returns the top K_l keywords with the 270

highest selection probability computed by the tagger. This is similar to (Saito et al., 2020a); they use the number of "guiding words" to control summary length, which can also be viewed as an instantiation of the CTRLSUM framework.

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Beyond above, more complicated logic may be incorporated into g_{control} to perform other control tasks - for instance, to achieve topic control, it is possible to select keywords that are related to a given topic word using external knowledge sources. Similar approach has been used in (Tan et al., 2020) for aspect-controlled summarization, whereas they require a specifically tailored training procedure. We leave such exploration as future work.

3.2 Combining Keywords and Prompts

Inspired by recent advances on prompting pretrained models (Liu et al., 2021), we explore to combine keywords and prompts to perform several different controllable summarization tasks.³ Specifically, we design a guiding token sequence for each task, which is used as *both* the keyword input and the decoder prompt. In this case the control signal c takes the task as a value, and $g_{\text{control}}(\mathbf{x}, \mathbf{c})$ outputs

³In this paper "prompt" is defined as the decoder prefix at test time, while prompt and keyword can be viewed as unified "prompt" which are both guiding tokens with different usage.

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294the guiding sequence directly given the task c for295any x. While traditional summarization systems296are able to utilize prompts to constrain and control297the generation as well, we emphasize that the com-298bination with keywords turns out to be important –299keywords help focus on control-related content on300the encoder side, compared to prompt alone which301only impacts the decoder, thus "keyword+prompt"302yields the best performance as we will show in §4.4.303Next we define and describe three control tasks that304we study in this paper.

Summarizing paper contributions: Existing datasets about scientific papers such as 306 307 arXiv (Cohan et al., 2018) collect paper abstracts as the summaries, which often include extra background context and lack detailed contribution descriptions for the associated paper. In many cases, readers would benefit from an explicit list 311 of contributions to understand the novelty of the 312 paper. For these cases, we propose g_{control} ="the 313 main contributions of this paper 314 are: (1)" as the keyword and prompt to generate 315 summary focused on contributions.

Summarizing invention purpose: Patent arti-317 cle summaries in existing datasets such as BIG-PATENT (Sharma et al., 2019) can be over-319 complicated, often covering core method details. 320 Yet for a non-technical reader it would be preferred 321 to provide a one-sentence summary on the purpose 322 of the invention while ignoring technical details. Thus we set g_{control} ="the purpose of the 324 present invention is" to yield a concise 325 summary on patent purpose.

Question-guided summarization: Human summarization can be constrained by questions (Kryściński et al., 2019) that require answers to be found in the summary. This points to a connection between summarization and reading comprehension that we further explore. We hypothesize that a summarization model may directly answer some questions about the article if guided properly. This suggests the possibility of subsuming reading comprehension as a form of summarization. To verify this, we set $g_{control}="Q: question text?$ A:" to produce reading comprehension behaviour.

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The guiding sequences we chose in this section are randomly-picked without any tuning, while in Appendix F we show that the performance is relatively robust across several human-written guiding sequences. Recent work on prompt engineering suggests that the guiding sequences may be further optimized automatically (Shin et al., 2020; Li and Liang, 2021), whereas they are not directly applicable without training data in the control tasks.

We note that keyword- and prompt-based control are complementary in practice – while prompts could ideally achieve any type of control, empirically they often do not work well for many aspects. For example, we found that using prompts alone (e.g. "a summary focused on [entity] is:") work poorly to control entities or length. Similarly, keyword alone struggles to perform paper contribution or patent purpose summarization. These empirical observations validate the importance of combining keyword and prompt for more generic controllable summarization. Please find quantitative analysis on this in Appendix E.

4 Experiments

In this section we test CTRLSUM quantitatively, and an extensive list of qualitative output examples in all the control tasks can be found in Appendix I.

4.1 Experimental Details

We perform experiments on three distinct-domain summarization datasets: CNN/Dailymail (CN-NDM) news articles (Hermann et al., 2015), arXiv scientific papers (Cohan et al., 2018), and BIG-PATENT patent articles (Sharma et al., 2019). The distribution $p(\mathbf{y}|\mathbf{x}, \mathbf{z})$ in CTRLSUM is our fine-tuned version of the pretrained BART_{LARGE} model (Lewis et al., 2019). Complete setup and training details can be found in Appendix A.1.

Evaluation: As emphasized in §2.2, we focus our evaluation on the five well-defined tasks. We measure ROUGE scores (Lin, 2004) and the recently proposed BERTScore (Zhang et al., 2020b) when ground-truth is available. In control task where we often do not have reference summaries, we (1) collect ground-truth summaries when possible, (2) examine whether summaries respect the control signal, or (3) resort to human evaluation.

4.2 Entity Control

Oracle entity setup: We first follow the evaluation in (Fan et al., 2018) to simulate a possible user preference by providing the model with oracle entities extracted from the ground-truth target. We note that this setting "cheats" by revealing the reference information and should be viewed as a sanity check experiment only to verify whether the model can utilize the input entities.

Table 2: ROUGE (1/2/L) scores with oracle entity or length signals from the reference summary. AK, OE, and OL denote automatic keywords, oracle entity, and oracle length respectively. LengthCode (†) is a length-control baseline from (Fan et al., 2018) re-implemented by us using BART.

Model	CNNDM	arXiv
	Unconstrained	
BART	44.24/21.25/41.06	45.16/17.36/40.55
CTRLSUM (AK)	45.65/22.35/42.50	46.91/18.02/42.14
	Constrained	
LengthCode [†]	43.44/21.10/40.35	45.91/17.33/41.38
CTRLSUM (OE)	48.75/25.98/45.42	-
CTRLSUM (OL)	46.26/22.60/43.10	47.58/18.33/42.79

Table 3: Entity control results on CNNDM assessing all possible entity inputs. Success rate is the fraction of summaries that mention the given entity, while factual correctness is from human annotators. The BART numbers are from unconstrained summaries. EntityCode (†) numbers are take from (Fan et al., 2018) which use a convolutional architecture.

Model	Succes	s Rate (%)	Factual Co	orrectness (%)		
Widdei	Lead-3 Full-article		Lead-3 Full-article Importa		Important	Unimportant
BART	61.4	29.0	98.0	_		
EntityCode [†]	61.2	33.8	-	-		
CTRLSUM	97.6	94.8	99.0	100.0		

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Any entity setup: We assess whether the model can deal with diverse preferences where every entity in the document is a possible input. To this end, we sample 100 test documents and repeatedly acquire every entity in the document to generate summaries, following Fan et al. (2018). Then we compute Success Rate, the fraction of requested entity actually occurring in the output. The results are reported in separation of whether the entity is from leading 3 sentences or the full article. To test if the summaries are factually consistent with the document, for each of the 100 documents we randomly sample one "important" entity that appears in the reference, and one "unimportant" entity that occurs neither in the reference nor the leading three source sentences. For each (article, summary) pair we ask 3 annotators from Amazon Mechanical Turk to make a binary decision as to whether the summary can be entailed from the article. We then take the majority vote and report the fraction of factually correct summaries. We evaluate on CNNDM only since many examples in arXiv and BIGPATENT do not have identifiable entities.

Results: In Table 2 we observe that the use of oracle entities helps boost the ROUGE-2 score⁴ compared with using automatic keywords, which indicates that CTRLSUM is able to take advantage

of the given entities. Table 3 shows the Success Rate and factual correctness evaluations. We include the numbers from (Fan et al., 2018) (*Entity-Code*) as a reference point which requires entity annotations during training time and is based on a convolutional seq2seq model.⁵ Remarkably, our model achieves a high success rate for both lead-3 and full-article entities reaching around 95%. Factual correctness scores from human annotators suggest that CTRLSUM is able to generate factually consistent summaries regardless of the importance of input entities, comparable to the BART baseline.

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4.3 Length Control

Oracle length setup: Similar to entity control, we examine the effect of oracle length signal from the reference for a sanity check.

Any length setup: To account for any possible length control signal (any length), we sample 1000 test documents and iterate every length bucket to decode several summaries for each document, and we measure whether the generated summaries follow the specified length signal. Specifically, we compute the mean of absolute deviation (MAD) of the actual length bucket code l_{sys} of the decoded summary from the given length $l_{\rm ref}$, as $\frac{1}{N}\sum_{n}^{N}|l_{\text{sys}}^{(n)}-l_{\text{ref}}^{(n)}|$. We also report the Pearson Correlation Coefficient (PCC) between the input bucket code and actual bucket code. We compare with LengthCode (Fan et al., 2018), where the ground-truth length bucket code is prepended to the article at both training and test time. Experiments are conducted on CNNDM and arXiv.

Results: In Table 2 CTRLSUM with oracle length signals only presents relatively small gains over the automatic CTRLSUM baseline. This implies that oracle lengths only convey limited extra information to help generate the reference. Table 4 includes results for the any length setup. Length-Code fails to consistently improve over BART with oracle length signals. Moreover, we find that LengthCode almost ignores the length signal with PCC close to 0, as shown in Table 4. This is possibly because that length code would be less helpful and thus more likely to be ignored when the model grows stronger through pretraining.⁶ In contrast, CTRLSUM with length-guided keywords achieves

⁵Please see Appendix D for ablation results on model architectures which indicates that the transformer architecture (Vaswani et al., 2017) is crucial for a high success rate.

⁴BERTScore results for Table 2 are in Appendix C.

⁶We verified that LengthCode using the BART model but randomly initialized weights obtained a high PCC score.

Table 4: Length control results assessing all possible length inputs. AK denotes automatic keywords. MAD measures the deviation of output length from the control length, while PCC represents the correlation between the them. LengthCode (Fan et al., 2018) is reimplemented by us using BART.

Model	CNN	DM	arXiv						
Model	$MAD\downarrow$	$MAD \downarrow PCC \uparrow$		PCC \uparrow					
Unconstrained									
BART	1.20	0.00	1.08	0.00					
CTRLSUM (AK)	1.25	0.00	0.98	0.00					
	Constrai	ned							
LengthCode	1.17	-0.02	1.06	0.00					
CTRLSUM (+length)	0.87	0.53	0.69	0.48					

high PCC between control signal and the actual output length, and is able to reduce the length deviation MAD compared to unconstrained baselines.

4.4 Contribution and Purpose Control

Contribution Summarization Setup: No existing dataset evaluates contribution summarization of scientific papers. However, researchers often summarize the bullet contributions of their paper in the Introduction section, which inspires us to extract them as the reference summary. Therefore, we resort to the entire arXiv database,⁷ and download all the papers whose first submission date is within the first six months of 2019⁸ that gives us 67K papers. We extract the Introduction section and bullet contributions with regular expression and filter out the ones that fail. The contributions are used as the reference and the Introduction section after removing the contribution claims is used as the source article. This procedure leads to 1018 test examples. We test the model trained on arXiv.

Purpose Summarization Setup: To collect a test dataset that features one-sentence invention purpose summaries, we sample 1000 test examples from BIGPATENT and present their reference summaries to human annotators from Amazon Mechanical Turk. For each example we ask one annotator to select the sentence that convey the invention purpose. We provide the option that the purpose cannot be identified. After filtering out the invalid examples, we collect 763 examples as our test data. We test the model trained on BIGPATENT.⁹

Results: Table 5 shows the results. Compared with BART which uses the prompt alone, CTRL-SUM achieves superior performance in most cases

Table 5: Summarization performance on paper contributions and invention purposes. P denotes prompts while K denotes keywords. BART baseline uses prompts while CTRLSUM use the same guiding tokens as both keywords and prompts.

Model	ROUGE-1/2/L	BERTScore (P/R/F1)
	Contribution	
BART (P)	43.84/17.46/25.89	0.119/0.142/0.130
CTRLSUM (P+K)	43.88/18.17/27.79	0.179/0.098/ 0.138
	Patent Purpose	
BART (P)	29.05/ 11.80 /22.50	0.016/0.236/0.107
CTRLSUM (P+K)	33.64 /11.37/ 24.24	0.180/0.152/ 0.165

Table 6: F1 scores on the dev set of NewsQA and SQuAD. P, K, FT denote prompt, keyword, and fine-tuning respectively.

Model	NewsQA	SQuAD v1.1
Supervised	d	
SpanBERT (Joshi et al., 2020)	73.0	94.6
MatchLSTM (Wang and Jiang, 2017)	49.6	70.0
Zero-Sho	t	
GPT2-Large (w/o FT, P)	24.9	23.5
BART (w/o FT, P)	8.2	15.8
BART (FT on CNNDM, P)	32.6	41.7
CTRLSUM (FT on CNNDM, P+K)	48.2	59.6

through using the guiding text as *both* the keywords and prompt. We further report the precision (P) and recall (R) scores in BERTScore besides F1. We observe that BART with prompt alone over-generates a full summary with low precision scores especially when a concise summary is desired in patent purpose summarization. CTRLSUM is able to mitigate this issue through keywords which guide the encoding process of encoder which prompts do not.

4.5 Question-Guided Summarization

Setup: We directly test question-guided summarization on reading comprehension benchmarks in a zero-shot fashion. Specifically, we evaluate the CNNDM models on in-domain NewsQA (Trischler et al., 2017) and out-of-domain SQuAD 1.1 (Rajpurkar et al., 2016) respectively.¹⁰ In addition to the BART baseline, we also include the performance from GPT2 language model (Radford et al., 2019) (without fine-tuning) as a reference point. Both BART and GPT2 use prompt alone to decode. We report F1 scores on the dev set.

Results: BART is pretrained with a denoising task to denoise the source, and unsurprisingly obtains poor results in the zero-shot setting, as shown in Table 6. Interestingly, however, BART fine-tuned on a summarization task – without seeing any question-answer pairs – is able to im-

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⁷https://www.kaggle.com/Cornell-University/arxiv

⁸The arXiv dataset used to train CTRLSUM is collected before April 2018, thus there should be no data overlap between the training data and our contribution test data.

⁹Both the contribution and purpose test data will be released after the review period for future research.

¹⁰Some NewsQA test articles are in the CNNDM training dataset, yet we still view it as a unsupervised setting since our model never sees questions or answers during training.

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Table 7: Unconstrained summarization performance. Automatic keywords are from the sequence tagger, while oracle keywords are obtained utilizing the gold summaries. We report the oracle performance for a reference point. BS denotes BERTScore.

Model	CNNDM		arXiv		BIGPATENT		
	ROUGE-1/2/L	BS	ROUGE-1/2/L	BS	ROUGE-1/2/L	BS	
CTRLSUM (Oracle Keywords)	64.65/40.42/60.92	0.555	56.08/25.31/50.23	0.268	55.19/26.62/47.10	0.291	
BART (Lewis et al., 2019)	44.24/21.25/41.06	0.336	45.16/17.36/40.55	0.164	45.83/19.53/39.47	0.187	
PEGASUS (Zhang et al., 2020a)	44.17/21.47/41.11	-	44.70/17.27/25.80	-	53.63/33.16/42.25	-	
CTRLSUM (Automatic Keywords)	45.65/22.35/42.50	0.363	46.91/18.02/42.14	0.169	45.80/18.68/39.06	0.188	

Table 8: Human evaluation on entity control and purpose control experiments (scale 1-5, higher is better). Control accuracy (CA) and control relevance (CR) are reported. A score significantly different (acc. to the Welch Two Sample t-test, with p < 0.05) than CTRLSUM is denoted by *.

Model Important Entity CA CR	Important Entity		Unin	portant Entity	Purpose		
	CA	CR	CA	CR			
CTRLSUM	3.5	4.2	4.0	4.0	4.0	3.7	
BART	3.8	3.7*	1.3*	1.2*	4.0	3.0*	

prove the F1 scores by 24.4 and 25.9 points on NewsQA and SQuAD respectively. Moreover, CTRLSUM with keywords and prompts further boost F1 by 15.6 and 17.9 points, approaching the supervised MatchLSTM score on NewsQA. Such results suggest that summarization might be a suitable transfer task for abstractive reading comprehension, which we leave for future work to explore.

4.6 Unconstrained Summarization

Table 7 shows the unconstrained summarization performance without any user input, where CTRL-SUM uses the automatically extracted keywords from the tagger as described in §2.3. On CNNDM and arXiv datasets CTRLSUM (Automatic Keywords) decently outperforms the strong BART and PEGASUS baselines. It also performs comparably to BART on BIGPATENT in terms of BERTScore, though with an inferior ROUGE-2 score. There is a performance gap between BART-based models and PEGASUS on BIGPATENT, possibly due to the inherent difference between BART and PEGASUS.

4.7 Human Evaluation

We present human evaluation results on controllable summarization tasks, but we refer the readers to Appendix B for human evaluation on unconstrained summarization. Full setup details are in Appendix A.2. We focus on entity and purpose control. Specifically, we inform the annotators our intent (which is to obtain summaries focused on a specific entity or purpose of patent), then we ask them to provide scores over two dimensions: (1) Control Accuracy (CA): whether the summary contains accurate main information of the intent, and (2) Control Relevance (CR): how the summary is

relevant to the control intent overall – a summary that contains redundant contents that are unrelated to the intent will be penalized. Results including significance tests are shown in Table 8. The control accuracy for important entity control and purpose control are comparable between BART and CTRLSUM without significant difference (*p*-value > 0.05), while CTRLSUM shows better control relevance overall by focusing on the desired information. Also, BART fails to generate unimportantentity-related summaries with poor scores.

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5 **Related Work**

Previous work on controllable summarization often collects control codes as supervision (Fan et al., 2018; Liu et al., 2018). For example, Fan et al. (2018) utilizes an entity tagger tool to identify entities from the reference summary, and input those entities together with the document during training for entity control. They also study length control by training a separate model with length code as the additional input. Other work mostly focus on a specific control task in limited settings (Saito et al., 2020a; Tan et al., 2020; Suhara et al., 2020), yet CTRLsum aims for a more generic framework.

Keyword-guided summarization has been applied in different contexts or with distinct motivations. Keywords are used as extra input to improve unconstrained summarization or reduce hallucinations (Gehrmann et al., 2018; Li et al., 2018; Saito et al., 2020b; Elsahar et al., 2020; Dou et al., 2021). CTRLsum is also related to query-focused summarization (Dang, 2005; Wan et al., 2007; Xu and Lapata, 2020), these methods often work with datasets that provide annotated queries, and did not focus on generic control tasks.

6 Conclusion

In this paper we propose a generic framework, CTRLSUM, to perform generic controllable summarization through a set of keywords. Experiments on five different control aspects demonstrate the efficacy of our method.

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A Experimental Setup Details

A.1 General Setup

In this section we include additional experimental details left out in the main content due to space limitations. The CNNDM, arXiv, and BIGPATENT source documents have an average of 790, 6914, 3573 tokens respectively. For all of them the source documents are truncated to 1024 tokens and the target summaries are truncated to 256 tokens following (Zhang et al., 2020a). We fine-tune the pretrained BART_{LARGE} model in all our experiments. Specifically we use the bart.large checkpoint from fairseq (Ott et al., 2019). For all BARTbased summarization models, we fine-tune with learning rate 3e-5 and a polynomial learning rate decay schedule, the optimizer is Adam (Kingma and Ba, 2015) and batch size is 64. Our optimization scheme and hyperparameters follow the BART fine-tuning instructions in fairseq examples. We train the summarization models with 20k steps on CNNDM, 50k steps on arXiv, and 300k steps on BIGPATENT. We train the BERT tagger with learning rate 5e-5, Adam optimizer, and batch size of 128 on all datasets. Similar to summarization models, the tagger is trained with 20k, 50k, and 300k steps on CNNDM, arXiv, and BIGPATENT respectively. Also, we adopt a sliding window approach so that the BERT-based tagger is able to handle sequences that are longer than 512 tokens.¹¹ For both ROUGE and BERTScore evaluation, we report the F1 measure. We report the rescaled BERTScore, and the hash code is "roberta $- large_L17_no - idf_version =$ $0.3.6(hug_trans = 3.0.2) - rescaled"$.

> The automatic keyword tagger at test time is based on the pretrained $\text{BERT}_{\text{LARGE}}$ model (Devlin et al., 2018) fine-tuned as described in §2.3. Our summarization model implementation is based on the fairseq toolkit (Ott et al., 2019) and the automatic keyword extraction model is based on the HuggingFace Transformers library (Wolf et al., 2019).

> As mentioned in §2.3, we need three hyperparameters for automatic keywords extraction during inference when applicable – the number of pre-selected sentences n_s , the selection probability threshold ϵ , and the maximum number of keywords m_{max} . We select these hyperparameters for

each dataset based on the unconstrained summa-
rization ROUGE-2 score on validation dataset. The
summarization performance is robust to these hy-
perparameters in a reasonable range, as shown in
Appendix G. Specifically, we use $\{n_s = 10, \epsilon =$
$0.25, m_{\text{max}} = 30$ for CNNDM, $\{n_s = 10, \epsilon =$
$0.15, m_{\max} = 40\}$ for arXiv, and $\{n_s = 5, \epsilon =$
$0.15, m_{\max} = 30\}.$

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The order and sentence boundary of automatic keywords: for the automatic keywords extracted as in §2.3, the keyword sequence maintains the order of the keywords as they were in the source document, but we observe that the model often ignores this ordering information as it frequently differs between source and target summary.¹² We also separate keywords from different source sentences with the special token ("|"), though in preliminary experiments we find that the model generation is not sensitive to the sentence boundary information at test time. At inference time in applications where the sentence boundary is unknown, as in most of our control tasks, the "|" token is ignored in the experiments.

Invention Purpose Summarization: in the experiment of summarizing invention purpose on patent articles (§4.4), we examined whether the model would possibly copy source sentences through matching the prompts, we search strings in the form of "the purpose of [some words or phrases] is" among 763 test examples, and only 3 test articles are identified. This means the models are not generating by exactly matching prompts most of the time.

Question-Guided Summarization: In Table 6, the GPT2-Large has 774M parameters while the BART architecture (including CTRLSUM) has 406M parameters.

A note on the unconstrained summarization experiment: in our unconstrained summarization experiment (§4.6), we note that the BERT tagger gives CTRLSUM parameter advantages over the baselines, leading to a strictly unfair comparison. Such a (unfair) parameter advantage setting is also adopted in recent two-stage summarization work (Dou et al., 2021). We constrain the seq2seq model in CTRLSUM to be the same as the baseline to better directly observe the effect of keywords.

¹¹https://github.com/google-research/ bert/issues/66.

¹²The keywords occurrence in the summary do not usually follow their order in the source document, thus the model seems not to utilize the order information in practice as we observed empirically.

Also, given that unconstrained summarization is
not the main focus of this paper, we leave a strict
parameter-control experiment for future work.

GPU resources. All our models including the
summarization model and tagger are trained on 8
NVIDIA Tesla 16GB V100 GPUs using a dataparallel mechanism. The GPUs are provided by
Google Cloud.

A.2 Human Evaluation Setup

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Here we include details about human evaluationexperiments in §4.7.

Unconstrained Summarization. For uncon-934 strained summarization, we sample 100 examples for each dataset, and hire workers from Amazon Mechanical Turk to conduct evaluation. Each ex-937 938 ample is scored by 3 independent workers and the resulted total number of participants are 15, 20, 22 for CNNDM, arXiv, and BIGPATENT respectively. We set the price of tasks in a way that will guarantee each worker an hourly wage of at least \$12. The 942 median score of 3 workers is taken for each exam-943 ple, and average over all examples is reported. For CNNDM we provide article and summaries, while 945 for arXiv and BIGPATENT we provide reference 946 and summaries using the reference summary as a 947 surrogate for the source article. This is because the source patent documents or scientific papers are very long and hard to be read by non-expert 950 humans. For each aspect to be scored (i.e. factual 951 consistency, relevance, fluency, or coherence), we provide specific definitions of the aspect. Then for each example we present summaries from different 954 systems together in random order to be evaluated. 955 Note that different aspects are presented and scored separately since we noticed that the MTurkers tend 957 to mix different aspect concepts and fail to distinguish them during evaluation if the four aspects are 959 scored together.¹³ An example screenshot of the 960 instructions shown to the workers is demonstrated 961 in Figure 2. 962

Controlled Summarization. For controlled summarization, we sample 100 examples for each task, and summaries of each example from different systems are presented together in random order to the human annotator to be scored. For CNNDM

we provide the source article, the control entities, and summaries, while for BIGPATENT we provide reference and summaries using the reference summary as a surrogate for the source article. This is because the source patent documents are very long and hard to be read by non-expert humans. We did not evaluate contribution summarization since it is unrealistic to ask non-expert humans to judge contributions of many scientific papers from various domains. We tried to hire workers from Amazon Mechanical Turk first, but unfortunately we failed to obtain reliable results from them they often ignored the given control intent and tended to score the text as standard unconstrained summaries, which is reflected by very poor scores on unimportant-entity-focused summaries because these summaries do not focus on the important information of the article, even though we instructed them that the control intent is critical. We note that obtaining reliable human judgement for summarization from non-expert crowdsourcing workers remains an active research direction (Shapira et al., 2019), and it has been shown before that non-expert human judgement of summarization can exhibit poor correlation with expert judgement (Fabbri et al., 2020).

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To obtain more trustable human evaluation results for controlled summarization, we ask three our colleagues to evaluate through personal correspondence from the authors of this paper. They are able to follow the given control intent from our manual investigation and considered more reliable than the MTurkers. While we report important and unimportant entity control results separately, we abstract away the "important/unimportant" entity identities from human annotators and mix them for the annotators to score. Similar to the unconstrained setting, each example is evaluated by 3 annotators independently. Then we take the median score for each example and average over all examples to report the final score. Example screenshots of the instructions shown to the human annotators are demonstrated in Figure 3.

B Human Evaluation Results on Unconstrained Summarization

We follow (Grusky et al., 2018; Fabbri et al., 2020)1013to ask human annotators from Amazon Mechanical1014Turk to score summaries over four dimensions: (1)1015Factual Consistency (FAC): the summary should1016only contain statements that can be entailed by1017

¹³This is reflected by a high correlation of our initially received scores among different aspects in our manual investigation. For example, a system often received a low score in the fluency dimension if it does not perform well on other dimensions like factual consistency, even though the generations are very fluent justified by us and our colleagues.

Definitions Consistency: The rating measures whether the facts in the text are consistent with t Consider whether the summary does reproduce all facts accurately an		Definitions Fluency This rating measures the quality of individual sentences, are they well-written and grammatically correct. Consider the quality of individual sentences.
Article	Generated Texts	Generated Text
\${article}	Text A	Text A
	\${test_oracleout}	\${test_keywordout}
	Consistency 1 2 3 4 5	Fluency 1 2 3 4 5
	Summary B	Text B
	\${test_keywordout}	\$(test_bartout)
	Consistency 1 2 3 4 5	Fluency 1 2 3 4 5
	Summary C	Text C
	\${test_bartout}	\${test_oracleout}

(a) Factual Consistency Scoring.

(b) Fluency Scoring.

Figure 2: Example screenshots of the instructions shown to the human annotators for unconstrained summarization. "\${}" denotes placeholders of the source article or different system outputs (they are replaced by actual text when presented to the MTurkers). This figure shows the aspect of "consistency" and "fluency" for an example, while we have additional two aspects of "relevance" and "coherence" scored. Different aspects are scored separately so that the MTurkers can better distinguish them. As shown in this figure, the annotators are provided with specific definitions of the aspect. The "fluency" dimension is scored without providing the source article to remove the potential source confounder since the fluency of the generated text piece is unrelated to the source article. Different system outputs are presented in random order.

source text	Entity	summary 1	Control Accuracy1	Control Relevance1	summary 2	Control Accuracy2	Control Relevance2
"Source text"is a news article, and "summary 1" and "summary 2" are two summaries that aim to describle the important information related to the entity.							
Grading Rubric: Control Accuracy (1-5): Consider whether the summary contains the important information about the entity correctly Control Relevance (1-5): How the summary is related to the entity overall. Summaries that contain a lot of unrelated information should be penalized							
There is often a negative signer attracted with being a female who chooses to live alone, expectably in Syndrey attraction sources 3-9 years of allow Seven the sine of an origination of the seven and a negative site of the seven and the seven the seven and the seven are more likely to have a university degree that the attract seven that degree that the seven and the seven are more likely to have a university degree that the seven and the seven are more likely to have a university degree that the seven and the seven are more likely to have a university degree that the seven and the seven are more likely to have a university degree that the seven and the seven are more likely to have a university degree that the seven and the seven are more likely to have a university degree that the seven and the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that the seven are more likely to have a university degree that t		A quarter of Australians have opted for solo living and 70 percent of the women are more likely to have a university degree than the men. 39-year-old Julie Sweet has lived alone in Clovelly for five years and has never been happier.			A new study shows that a quarter of Australians have opted for solo living. 70 percent of the women are more likely to have a university degree than the men. The affluent suburks of Potts Point, Rushcutters Ray and Elizabeth Bay make up nearly 60 percent of solo households. This means that the women who live alone in the Eastern suburbs are fast becoming more successful than their male counterparts.		

(a) Entity Control on CNNDM.

source text	summary 1	Control Accuracy1	Control Relevance1	summary 2	Control Accuracy2	Control Relevance2
The "source text" column includes a description for a patent invention, and the "summary 1" and "summary 2" are two different text that aim to describle the purpose of the invention. The task for the humans is to read them and score (with 1,2,3,4,5) the quality of the purpose summaries.						
Grading Rubric: 1. Control Accuracy (1-5): whether the purpose description in the summary is accurate 2. Control Relevance (1-5): how the summary is related to "purpose of invention" overall. Summaries that contain extra unrelated-to-purpose information should be penalized on this score						
combining random selection of a pass number for printing each pixel of a printmask with application of various constraints. In combination with thore aspects of the invention, speed is optimized by depositing substantially a single drop of ink per pixel, thereby minimizing both the number of passes required to render all pixels completely and driving time. Image quality is optimized by maximizing the time and distance between deposition of individual ink come, thereby unimiprine and greaces. All before, and between the temperature and the temperature of the second secon	the purpose of the present invention is to provide a method and apparatus for generating a mask for printing an image on media . the method comprises the steps of . receiving image data .generating a first mask from themge data using a random number and a constraint ; and printing the first mask on media using a printer. the apparatus comprises an image source ; a mask generator ; and a printer coupled to the image source and the mask generator .			the purpose of the present invention is to provide an improved method and apparatus for generating masks for plotters .		

(b) Purpose Control on BIGPATENT.

Figure 3: Example screenshots of the instructions shown to the human annotators for controlled summarization. (a) shows the entity control evaluation example on CNNDM and (b) demonstrates the purpose control instructions on BIGPATENT. We present different system outputs in random order to remove order biases, which means that the generations in the same column ("summary1" or "summary2") may not be from the same system.

Table 9: Human evaluation scores (scale 1-5, higher is better) of unconstrained summarization performance. Evaluation Dimensions from left to right are: factual consistency (FAC), relevance (REL), fluency (FLU), coherence (COH). A score significantly different (according to the Welch Two Sample t-test, with p < 0.05) than CTRLsum (Automatic Keyword) is denoted by *.

Model	CNNDM FAC/REL/FLU/COH	arXiv FAC/REL/FLU/COH	BIGPATENT FAC/REL/FLU/COH
CTRLsum (Automatic Keyword)	4.6/4.6/4.1/4.1	4.1/4.3/4.1/4.1	4.2/4.2/4.0/4.1
BART	4.6/4.7/4.2/4.1	4.1/4.1*/3.9/4.0	4.2/4.3/4.1/4.0
CTRLsum (Oracle Keyword)	4.6/4.7/4.1/4.1	4.2/4.3/4.0/4.1	4.2/4.2/4.2*/4.1

Table 10: BERTScore with oracle entity or length signals from the reference summary. "CTRLsum (automatic)" represents our model using automatic keywords in an unconstrained setting. LengthCode (†) is a length-control baseline from (Fan et al., 2018) re-implemented by us using BART.

Model	CNNDM	arXiv
BART (Lewis et al., 2019)	0.336	0.164
CTRLsum (automatic)	0.363	0.169
LengthCode [†]	0.346	0.147
CTRLsum (oracle entity)	0.422	-
CTRLsum (oracle length)	0.365	0.173

the source document, (2) Relevance (REL): the summary should only contain *important* information of the source document, (3) Fluency (FLU): each sentence in the summary should be fluent, and (4) Coherence (COH): the summary should be well-structured and well-organized. More setup details are in Appendix A.2. Table 9 shows the results including significance tests. The quality of summaries from all systems on all dimensions is generally good with a score mostly higher than 4.0. However, most scores do not show significant difference from CTRLsum (Automatic Keyword), despite their very different ROUGE/BERTScore against the reference summaries. This implies that the summary quality from different systems powered by strong pretrained models like BART has become difficult to be clearly distinguished by nonexpert MTurkers.

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C Additional BERTScore Results

Table 10 shows the additional BERTScore results when oracle entity or length signals are used.

D Ablation Analysis of Entity Control

In Table 3 we observe that CTRLsum achieves a very high success rate ($\sim 95\%$) of entity control, compared to previous work (Fan et al., 2018) which can only succeed 61.2% and 33.8% of the time on lead-3 and full-article entities respectively. We perform ablation analysis to understand the important ingredients that contribute to the success of CTRLsum. We train CTRLsum with another 1047 two architectures in addition to BART: (1) convolu-1048 tional seq2seq (Gehring et al., 2017) with the same hyperparameters as in (Fan et al., 2018), and (2) 1050 transformer seq2seq with the same hyperparame-1051 ters as the base model in (Vaswani et al., 2017). Note that the transformer model is trained from 1053 scratch without pretraining. Results are shown in Table 11. CTRLsum parameterized with a weaker 1055 convolutional seq2seq architecture fails to depend 1056 on the keywords well with an over 40-point suc-1057 cess rate drop, yet the success rate of transformer seq2seq without pretraining only drops around 5 1059 points. This implies that the transformer seq2seq 1060 architecture is critical for CTRLsum to depend on the keywords well, while pretraining can further 1062 improves it.14 1063

Table 11: Entity control results on CNNDM. Success rate is the fraction of decoded summaries that actually mention the given entity.

Model	Success Rate (%)		
Model	Lead-3	Full-article	
BART (Lewis et al., 2019)	61.4	29.0	
Fan et al. (2018)	61.2	33.8	
CTRLsum (Conv Seq2Seq)	50.1	23.3	
CTRLsum (Transformer Seq2Seq)	92.6	88.3	
CTRLsum (BART)	97.6	94.8	

E Are Keywords and Prompts Complementary?

In this paper CTRLsum uses control tokens either as keywords alone (entity and length), or as both keywords and prompts (contribution, purpose, QA). Here we present further results when control tokens are used as prompts, keywords, or both for entity control, contribution control, and NewsQA tasks. Specifically for entity control, we use the control to-

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¹⁴For reference points, the ROUGE-1/2/L scores (with automatic keywords) of CTRLsum (Conv Seq2Seq) is 41.19/18.71/38.05 while CTRLsum (Transformer Seq2Seq) obtained 43.69/20.78/40.55.

Table 12: Ablation analysis on the role of keyword and prompt respectively. Entity success rate refers to the full article entity success rate. K denotes keyword while P denotes prompt.

Model	Entity	Contribution		NewsQA
widdei	Success Rate (%)	ROUGE-1/2/L	BERTScore	F1
CTRLsum (K)	94.8	39.96/12.74/22.68	0.088	15.5
CTRLsum (P)	17.6	43.82/18.12/27.64	0.133	26.3
CTRLsum (K+P)	12.6	43.88/18.17/27.79	0.138	48.2

kens "a summary focused on [entity] 1073 is:" for "prompt" and "prompt + keyword" vari-1074 ants.¹⁵ In this case success rate is computed exclud-1075 ing the prompt text. The control tokens for other 1076 settings are the same as previous experiments. Re-1077 sults are shown in Table 12, where keywords and 1078 prompts are of different importance for different 1079 tasks and are complementary in general. For exam-1080 ple, using prompts to control entities turns out to 1081 be difficult with a very low success rate - we find 1082 that the system fails to understand the prompt and produce summaries appropriately in most cases. 1084 However, prompts contribute the most to contribu-1085 1086 tion summarization with comparable performance with using prompts and keywords together, while 1087 removing prompts and using keywords alone suf-1088 fers from drastic performance drop to trigger the 1089 contribution. For NewsQA task, prompts and key-1090 words demonstrate mixing effectiveness - using 1091 1092 either of them alone experiences over 20 F1 points loss compared to using them together. 1093

F Sensitivity Analysis of Different Prompts

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In §4.4 and §4.5 we used randomly picked prompts without any tuning process, which represents a practical setting where we often do not have supervision for the specific control tasks at hand. This stands in contrast with recent prompt optimization methods which require a non-trivial amount of training examples to work well (Shin et al., 2020; Li and Liang, 2021; Perez et al., 2021). To analyze the model's sensitivity to different random prompts, we study contribution and purpose summarization with three randomly-picked prompts for each, and report the mean and standard deviation of ROUGE scores. Table 13 shows the three specific prompts we use, and Table 14 demonstrates the results. While we cannot iterate over all possible prompts, these results indicate that CTRLsum1111can work reasonably well with a reasonable set of1112simple, randomly-picked, human-written prompts.1113

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Table 13: The prompts we use for contribution and purpose summarization respectively. We use three different prompts for each to analyze model's sensitivity to different prompts.

Contribution	the main contributions of this paper are: (1) the contributions of this paper can be summarized as: (1) the contributions of this paper are: (1)
Purpose	the purpose of the present invention is the goal of the invention is the invention is designed with the goal of

G Robustness Analysis of Keywords Extraction Hyperparameters

Table 16 shows the ROUGE-2 scores of unconstrained summarization on the validation set with different keywords extraction hyperparameters. We use more fine-grained stride size to iterate the m_{max} hyperparameter for CNNDM since its source articles are usually shorter than arXiv and BIG-PATENT. As observed, the automatic summarization performance is relatively robust to these hyperparameters in a reasonable range.

H Results on Query-Based Summarization Datasets

The goal of CTRLsum is to perform controllable 1127 summarization tasks with standard summarization 1128 datasets resources, i.e. only using the source docu-1129 ment and summary during model training and vali-1130 dation. This differs from the research line of query-1131 focused summarization (Dang, 2005; Wan et al., 1132 2007; Xu and Lapata, 2020) which often deals with 1133 query-based datasets that contain (query, document, 1134 summary) triples. However, we think that it may be 1135 of interest to examine the performance of CTRL-1136 sum on query-based benchmarks as well. To this 1137 end, we follow (Xu and Lapata, 2021) and directly 1138 evaluate the pretrained, CNNDM CTRLsum model 1139 on the test set of query-based datasets without re-1140 training, in a zero-shot transfer setting. We note 1141 that this experiment does not belong to our focused 1142

¹⁵We tried several prompts variants, for example, QA style ones "Q: What happened to [entity]? A:" or "Q: What do we know about [entity]? A:". None of them lead to meaningful entity control.

Table 14: Summarization performance on contributions of papers and purpose of inventions. The BART baseline uses prompts (P) while CTRLsum uses both keywords (K) and prompts (P). We run with three different randomly-picked prompts and report the mean and standard deviation (in the bracket) of scores.

Model	Contribution ROUGE-1/2/L (std)	Patent Purpose ROUGE-1/2/L (std)
BART (P)	43.77 (0.12)/17.53 (0.12)/25.71 (0.17)	28.22 (0.77)/ 11.23 (0.55)/21.96 (0.49)
CTRLsum (K+P)	43.97 (0.09)/18.31 (0.12)/27.78 (0.07)	32.43 (1.49)/10.87 (0.69)/ 23.66 (0.96)

Table 15: Zero-shot performance on Debatepedia test set. All the baseline results are copied from (Xu and Lapata, 2021). LexRank and Gsum results are from extended version of the respective methods for this setting implemented by (Xu and Lapata, 2021). For CTRLsum we use the pretrained model from CNNDM and directly run the evaluation.

Model	ROUGE-1/2/L	
Lead	18.1/5.6/15.9	
LexRank (Erkan and Radev, 2004)	17.4/5.3/15.1	
Supervised (Abstractive)		
DDA (Laskar et al., 2020)	7.4/2.8/7.2	
BertAbs+Rank (Abdullah and Chali, 2020)	19.2/10.6/17.9	
BertAbs+Concat (Laskar et al., 2020)	26.4/11.9/25.1	
Zero-shot (Abstractive)		
BertAbs (Liu and Lapata, 2019)	13.3/2.8/2.8	
Bart (Lewis et al., 2019)	21.4/6.3/18.4	
GSum + Query (Dou et al., 2021)	21.2/6.2/18.2	
LaQSum (Xu and Lapata, 2021)	23.5/7.2/20.6	
CTRLSUM	22.7/7.2/20.0	

scope of control tasks and is presented mainly forcompleteness purpose.

Query-based summarization datasets are scarce 1145 mainly due to the expensive annotation efforts 1146 of queries. The most commonly used ones are 1147 probably the DUC corpus (Dang, 2005) which is 1148 a multi-document summarization dataset. Given 1149 that CTRLsum focuses on single-document summa-1150 rization and adapting pretrained single-document 1151 summarization models to multi-document cases re-1152 quires additional pipelines (Baumel et al., 2018), 1153 1154 we follow (Xu and Lapata, 2021) and perform experiments on a single-document, query-based sum-1155 marization benchmark Debatepedia (Nema et al., 1156 2017). Debatepedia is a summarization dataset 1157 with question queries to guide the summary. We 1158 use the CTRLsum model trained on CNNDM to 1159 predict summaries for the Debatepedia test set. We 1160 use the provided queries as the test keywords \mathbf{z}_{test} in 1161 CTRLsum. We compare with LexRank which ex-1162 tracts sentences that are similar to the queries as the 1163 summary. We also include supervised numbers as 1164 a reference point and state-of-the-art (query-based) 1165 baselines in a zero-shot setting. 1166

As shown in Table 15, CTRLSUM is **Results:** 1167 able to outperform BART and strong query-guided 1168 system GSum (Dou et al., 2021), which implies 1169 that CTRLSUM is able to generalize relatively well 1170 to query-based benchmarks. CTRLSUM underper-1171 forms recently proposed LaQSum (Xu and Lapata, 1172 2021) on Rouge-1 and Rouge-L which learns la-1173 tent queries. We emphasize that LaQSum does 1174 not focus on controllable summarization tasks and 1175 is not able to perform the series of control tasks 1176 present in this paper without our control function. 1177 Compared to supervised baselines that utilize pre-1178 trained models and the provided queries in Debape-1179 dia during training, CTRLSUM's performance lags 1180 behind which reflects the gap between the training 1181 keywords (which is extracted from the reference 1182 summary) and the test-time keywords (which is pro-1183 vided as question queries along with the dataset). 1184

Table 16: ROUGE-2 scores of unconstrained summarization on the validation set with different keywords extraction hyperparameters.

Model	CNNDM	arXiv	BIGPATENT
$\epsilon = 0.10, n_s = 5, m_{\max} = 25$	22.82	_	_
$\epsilon = 0.10, n_s = 5, m_{\max} = 30$	22.71	17.81	18.60
$\epsilon = 0.10, n_s = 5, m_{\max} = 35$	22.54	_	_
$\epsilon = 0.10, n_s = 5, m_{\max} = 40$	_	17.96	18.35
$\epsilon = 0.10, n_s = 10, m_{\max} = 25$	22.83	_	_
$\epsilon = 0.10, n_s = 10, m_{\max} = 30$	22.67	17.99	18.61
$\epsilon = 0.10, n_s = 10, m_{\max} = 35$	22.44	_	—
$\epsilon = 0.10, n_s = 10, m_{\max} = 40$	-	18.03	18.04
$\epsilon = 0.15, n_s = 5, m_{\max} = 25$	22.85	-	-
$\epsilon = 0.15, n_s = 5, m_{\max} = 30$	22.79	17.80	18.79
$\epsilon = 0.15, n_s = 5, m_{\max} = 35$	22.71	_	_
$\epsilon=0.15, n_s=5, m_{\rm max}=40$	-	17.95	18.76
$\epsilon = 0.15, n_s = 10, m_{\max} = 25$	22.85	_	_
$\epsilon = 0.15, n_s = 10, m_{\max} = 30$	22.77	17.99	18.76
$\epsilon = 0.15, n_s = 10, m_{\max} = 35$	22.41	_	_
$\epsilon = 0.15, n_s = 10, m_{\max} = 40$	_	18.05	18.62
$\epsilon = 0.20, n_s = 5, m_{\max} = 25$	22.86	—	_
$\epsilon = 0.20, n_s = 5, m_{\max} = 30$	22.87	17.71	18.77
$\epsilon = 0.20, n_s = 5, m_{\rm max} = 35$	22.89	—	—
$\epsilon = 0.20, n_s = 5, m_{\max} = 40$	-	17.88	18.71
$\epsilon = 0.20, n_s = 10, m_{\max} = 25$	22.87	_	—
$\epsilon = 0.20, n_s = 10, m_{\max} = 30$	22.85	17.88	18.77
$\epsilon = 0.20, n_s = 10, m_{\max} = 35$	22.84	_	—
$\epsilon = 0.20, n_s = 10, m_{\max} = 40$	-	17.98	18.73
$\epsilon = 0.25, n_s = 5, m_{\max} = 25$	22.84	_	_
$\epsilon = 0.25, n_s = 5, m_{\max} = 30$	22.88	17.57	18.67
$\epsilon = 0.25, n_s = 5, m_{\max} = 35$	22.91	_	_
$\epsilon = 0.25, n_s = 5, m_{\max} = 40$	_	17.71	18.66
$\epsilon = 0.25, n_s = 10, m_{\max} = 25$	22.90	_	_
$\epsilon = 0.25, n_s = 10, m_{\max} = 30$	22.95	17.76	18.72
$\epsilon = 0.25, n_s = 10, m_{\max} = 35$	22.95	-	-
$\epsilon = 0.25, n_s = 10, m_{\max} = 40$	-	17.84	18.70
$\epsilon = 0.30, n_s = 5, m_{\max} = 25$	22.58	_	-
$\epsilon = 0.30, n_s = 5, m_{\max} = 30$	22.62	17.24	18.53
$\epsilon = 0.30, n_s = 5, m_{\max} = 35$	22.63	-	-
$\epsilon = 0.30, n_s = 5, m_{\max} = 40$	-	17.32	18.52
$\epsilon = 0.30, n_s = 10, m_{\max} = 25$	22.65	-	-
$\epsilon = 0.30, n_s = 10, m_{\max} = 30$	22.70	17.38	18.55
$\epsilon = 0.30, n_s = 10, m_{\max} = 35$	22.70	-	-
$\epsilon = 0.30, n_s = 10, m_{\max} = 40$	-	17.44	18.55

I Random Output Examples

In this section, we randomly sample test examples and show the source aticle, reference summary, and the model output from CTRLsum for each control aspect. 1187

I.1 Entity Control

For entity control, we randomly sample 3 articles from CNNDM and for each article we randomly select 5 entites as keywords to show the model output.

Table 17: Random Entity Control Examples

Article	Americans on the United States' no-fly list will now be privy to information about why they have been banned from commercial flights and be given the opportunity to dispute their status, according to court documents filed by the Justice Department this week. The revised policy comes in response to a June ruling by a federal judge that said the old process was in violation of the Fifth Amendment's guarantee of due process. The decision was part of an American Civil Liberties Union lawsuit brought on behalf of 13 Americans on the list. But the ACLU isn't satisfied with the government's new policy, outlined in documents filed Monday in federal courts in Oregon (PDF) and Virginia (PDF). "After years of fighting in court for complete secrecy and losing, it's good that the government is finally now going to tell people of their status on the No Fly List," said Hina Shamsi, director of the ACLU National Security Project and the lead attorney on the case, in a statement. "Unfortunately, we've found that the government's new redress process falls far short of constitutional requirements because it denies our clients meaningful notice, evidence, and a hearing. The government had an opportunity to come up with a fair process but failed, so we're challenging it in court again." People on the no-fly list, managed by the FBI's Terrorist Screening Center, are prohibited from boarding a commercial flight for travel into or out of the United States. The number of people on the list is classified. An official with knowledge of the government's figures told CNN in 2012 that the list contained about 21,000 names, including about 500 Americans. Before the change, American citizens and permanent residents who inquired with the government is making enhancements to the Department of Homeland Security Traveler Redress Inquiry Program (DHS TRIP) to provide additional transparency and process for U.S. citizens and lawful permanent residents who have been denied boarding on a commercial aircraft because they are on the No Fly List, "the D
Reference Summary	Americans on the no-fly list will now get info about why they've been banned from flights. ACLU says the policy still denies "meaningful notice, evidence, and a hearing"
	the Fifth Amendment's : The new policy is in response to a June ruling that said the old process was in violation of the Fifth Amendment's guarantee of due process.
	Civil Liberties Union: The American Civil Liberties Union says it's not satisfied with the government's new pol- icy.
CTRLsum	the Transportation Security Administration : The Transportation Security Administration will provide travelers with more information about their status. A June ruling had called for passengers on the list to be given the opportunity to dispute their status before a judge.
	Virginia: The ACLU is challenging the new policy in federal courts in Oregon and Virginia.
	FBI: People on the no-fly list, managed by the FBI's Terrorist Screening Center, are prohibited from boarding commercial flights. The ACLU isn't satisfied with the government's new policy.

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Table 18: Random Entity Control Examples

Article	Point guard Stephen Curry nearly single-handedly outscored New Orleans with 11 first-quarter points as the Warriors built a 15-point lead and rolled to victory in Game One of their Western Conference first-round series. Game Two in the best-of-seven series is scheduled for Monday night in Oakland. Golden State, the top seed in the West, picked up right where it left off in the regular season, recording a 19th straight home win and 40th in 42 games this year. Stephen Curry scored a stunning 34 points for the Golden State Warriors in there play-off game. The Warriors did it by taking a 25-point lead into the final minute of the third quarter, then holding on, 'We missed a lot of free throws, which made it a lot closer than it needed to be,' coach Steve Kerr said. 'But in the playoffs you've just got to get it done somehow. We're up 1-0. That's where we want to be.' Curry led the Warriors with 34 points, hitting 13 of 25 shots and four three-pointers. All five Golden State starters scored in double figures. Guard Klay Thompson complemented Curry with 21 points, while power forward Draymond Green (15 points, 12 rebounds) and center Andrew Bogut (12 points, 14 rebounds) recorded double-doubles. The point guard has been in spectacular form as he looks to lead the Warriors to the NBA glory. Curry celebrates after scoring a three-pointer on his way to scoring 11 first quarter points. New Orleans power forward Anthony Davis scored a game-high 35 points, 20 in the fourth quarter. He also blocked four shots and grabbed seven rebounds. Small forward Quincy Pondexter, assigned to cover Curry most of the game, added 20 points for the Pelicans Visiting Washington outscored Toronto 11-4 in the overtime in winning the opener of their first-round playoff series. Washington traveled to Toronto and came away with the victory with a confident overtime performance. Forward Paul Pierce started the overtime with a three-pointer and scored 20 points for the Pelicans Visiting Washington cutscored Toronto 11-6 points for Washingto
Reference Summary	Stephen Curry scored 34 points for Golden State against New Orleans. The Californian-based team defeated the Pelicans 106-99. Washington Wizards outscored the Toronto Raptors 11-4 in overtime. Paul Pierce led the scoring with 20 points for the Wizards .
CTRLsum	 Stephen Curry: Stephen Curry scored 34 points as Golden State Warriors beat New Orleans Pelicans. Washington Wizards beat Toronto Raptors in overtime in play-off game. Oakland: Stephen Curry scored 34 points for the Golden State Warriors. Game Two in the best-of-seven series is scheduled for Monday night in Oakland. Steve Kerr: Stephen Curry scored 34 points as the Warriors beat the Pelicans. Coach Steve Kerr said: 'We missed a lot of free throws, which made it a lot closer than it needed to be' Klay Thompson: Stephen Curry scored 34 points as the Warriors beat the Pelicans. Klay Thompson and Draymond Green also scored in double figures . Andrew Bogut: Stephen Curry scored 34 points as the Warriors beat the Pelicans. Draymond Green and Andrew Bogut both recorded double-doubles.

Table 19: Random Entity Control Examples

Article	It's the ultimate treat for Benedict Cumberbatch fans and stands an imposing 6ft tall - just like the man himself. But shoppers at London's Westfield Stratford City shopping centre looked more than a little surprised to discover a chocolate sculpture of Benedict Cumberbatch in their midst. One lady was spotted cautiously approaching the edible artwork before quickly backing off, while another couldn't quite hide their smile of surprise. Scroll down for video . Finishing touches: The sculpture is readied for its big unveiling at Westfield Stratford City shopping centre. Oh dear: Reaction to the sculpture was mixed, with some shoppers bursting into laughter. Even less impressed was the shopper who stood stony-faced in front of the creation for several moments, while another cougaged their tails as they gave it a thorough sniffing down. The artwork, which has been given pride of place in the shopping mall's atrium, was commissioned by UKTV to mark celebrate its screening of the third series of Sherlock. It took a crew of eight people to complete the sculpture, which took over 250 man hours to create and weighs 40kg. Does it look like me? Benedict Cumberbatch strikes a pose with James Corden during an Oscars party. Mixed reaction: A pair of police sniffer dogs loved the sculpture but shoppers looked baffled. Hilarious: A lady bursts into laughter after spotting the 6ft homage to Mr Cumberbatch. Not amused: A shopper looks thoroughly unimpressed as she contemplates the artwork. Luckily for Cumberbatch, who usually enjoys a considerably more complimentary response to projects he's involved in, the piece will only be in residence temporarily. The 38-year-old actor, who is currently expecting his first child with wife Sophie Hunter, 37, in't the only famous face to have found himself the subject of an edible artwork. In the run up to the release of 50 Shades of Grey, bakers created not one but two ft gateaux paying homage to Jamie Dornan. One depicted the actor in the grey suit beloved of his 50 Shades character Christ
Reference Summary	A 6ft chocolate sculpture of Benedict Cumberbatch has been unveiled. Toothsome statue has been placed inside a London shopping centre. But shoppers reactions to the creations were decidedly unenthusiastic. One woman glared at it while others just looked thoroughly baffled . It did manage to win the approval of pair of police sniffer dogs. It weighs 40kg and took eight people 250 man hours to create . Other celebrities to get culinary tributes include Jennifer Lawrence. Her 6ft cake won an award - and the 24-year-old's approval. Actor Kevin Bacon has also been immortalised - in bacon .
	Westfield: The sculpture was unveiled at London's Westfield Stratford City shopping centre.
	James Corden: Benedict Cumberbatch is not the only famous face to be immortalised in chocolate. Jennifer Lawrence and James Corden have also been turned into cakes .
CTRLsum	Sophie Hunter: Actor, 38, is expecting his first child with wife Sophie Hunter.
	Jamie Dornan: Shoppers at London's Westfield Stratford City shopping centre looked baffled by the sculpture. Other famous faces to be immortalised in edible art include Jamie Dornan and Jennifer Lawrence.
	Hunger Games : The sculpture was commissioned by UKTV to mark the screening of Sherlock. It follows in the footsteps of other A-listers such as Jamie Dornan and Jennifer Lawrence, who have been immortalised in cakes. Lawrence's Hunger Games cake won an award at a show last year.

I.2 Paired Entity Control

The entity control experiments in this paper only consider one entity as the control signal, here we show	1192
examples inputting paired entities. Specifically, we are interested in the output when one of the paired	1193
entities is important and the other is unimportant. Therefore, we sample 3 articles from CNNDM and	1194
for each article we randomly select one important entity and one unimportant entity as paired keywords	1195
to show the model output. We repeat this sampling five times for each article to obtain five different	1196
summaries.	1197

Table 20: Random Paired Entity Control Examples

Article	A former U.S. Army enlistee who posted on Facebook about "the adrenaline rush" of dying in jihad was arrested Friday an charged with trying to detonate a car bomb at Fort Riley military base in Kansas, authorities said. A second man, who allegeld knew about the bomb plot but dird in't call authorities, was charged with failing to report a felony. John T. Booker T. of Topekk an American citizen also known as Mohammed Abdullah Hassan, was taken into custody near Manhattan, Kansas, in a vai that contained what he thought was a bomb, the criminal complaint said. The "bomb" had actually been put together by two confidential informants with nonexplosive materials, the complaint said. Fort Riley's security was never breached and no peoph were in danger, the U.S. Justice Department said in a press release. Booker enlisted in the Army last year and was due to shi out to basic training April 7, 2014, said Army spokesman Wayne Hall. The criminal complaint said the FBI questioned hin March 24, 2014 about comments posted on Facebook, such as, "Getting ready to be killed in jihad is a HUGE adrenaline rusl I am so nervous. NOT because I'm scare to die but I am eager to meet my lord." Booker waived his Miranda rights and tolt the agents he enlisted to commit an insider attack against American soldiers like Maj. Nidal Hassan had done at Fort Hoor Texas, the complaint said. Hassan opened fire in a building in November 2009, killing 13 people and wounding more than 33 His enlistment was terminated March 24, 2014, at the request of Army Criminal Investigation Command, Hall said. Booke began communicating with a confidential informant later in 2014, the complaint said, and often talked about his plans to engag in violent jihad in support of ISIS. He and the informant watched ISIS videos together, the complaint said, and Booker talke about how he wanted to go to Iraq and turn his weapon on American soldiers when ordered to shoot the enemy. On March 2 Booker said he believed ISIS wanted him to commit a truck bombing in the United Sta
Reference Summary	Alexander Blair, 28, of Topeka accused of knowing about bomb plot but not contacting authorities. Fort Riley's security wa never breached and the device was "inert" and not a threat, authorities say. John T. Booker Jr., 20, of Topeka had acquired bomb parts and made a propaganda video, the Justice Department says.
	Fort Riley ISIS: John T. Booker Jr. is charged with trying to detonate a car bomb at Fort Riley. A second man, who allegedl knew about the bomb plot but didn't call authorities, is charged. Booker said he believed ISIS wanted him to commit a true bombing in the United States.
	John T. Booker Jr. FBI: John T. Booker Jr. enlisted in the Army last year and was due to ship out to basic trainin, April 7, 2014. The FBI questioned him March 24, 2014 about comments posted on Facebook.
CTRLsum	John T. Booker Jr. Infantry Division: John T. Booker Jr. enlisted in the Army last year and was due to ship out t basic training April 7, 2014. He planned to detonate a car bomb at Fort Riley, a large Army base that's home to the 1st Infantr Division.
	Justice Department Islamic State of Iraq: The Justice Department says Fort Riley's security was never breacher and no people were in danger. John T. Booker Jr. is charged with attempting to use a weapon of mass destruction. He is als charged with trying to provide material support to the Islamic State of Iraq.
	Alexander Blair ISIS: John T. Booker Jr., 28, charged with trying to detonate a car bomb at Fort Riley. Alexander E. Blair, 28, is charged with failing to report a felony. Booker said he wanted to commit an insider attack against America soldiers in support of ISIS.

Table 21: Random Paired Entity Control Examples

Article	Ander Herrera has caught the eye in recent weeks after cementing a spot in Manchester United's starting line-up but it appears he does not actually have a clear sight at goal. Herrera netted a brace against Aston Villa on Saturday afternoon without actually looking at Brad Guzan's goal as his eyes appeared to be shut when making contact with the ball. In fact, six of Herrera's seven goals have been scored without him even having to glimpse at either the ball or the opposition's net. Manchester United star Ander Herrera scores his side's opening goal against Aston Villa with his eyes shut. The Spanish midfielder appears to have his eyes closed as he strikes at Brad Guzan's goal . Aston Villa and England midfielder Fabian Delph attempts to block Herrera's left-footed shot. His eyes were wide open when he struck an impressive first-time shot against Yeovil in the third round of the FA Cup back in January. However his double against Aston Villa and his goals against Queens Park Rangers, Leicester, Preston and Swansea all came without Manchester United's summer signing having to make eye contact with the ball. Herrera appears to have a history of shooting with his eyes closed as the image of his goal for Spain's Under 20 side back in 2009 shows. Herrera hits the back of the net while representing Spain's Under 20 side against Tahiti . The former Athletic Bilbao joined the Red Devils for £29million in the summer due to his vision and creativity in midfield . Louis van Gaal heaped praise on the 25-year-old's shooting ability after Saturday's Premier League encounter at Old Trafford, by stating: 'He has a very good kicking technique and he should be more composed,' Van Gaal said to MUTV. 'I said to him 'you have to control the ball before you shoot'. 'I said that to him again in yesterday's training session, and then I kissed him at half-time because he had controlled the ball for the first time in his life and then he shot.' . Herrera's eyes were also shut for Manchester United's third goal against Aston Villa on Satu
Reference Summary	Ander Herrera has scored seven goals for Man United since joining in June. Herrera's eyes have been shut when striking the ball for six of his goals. His superb strike against Yeovil Town has been only goal with eyes open. Herrera netted a brace in Manchester United's 3-1 win over Aston Villa.
	Ander Herrera Van Gaal: Ander Herrera has scored six of his seven goals without looking at the ball. Louis van Gaal has praised the Spanish midfielder's shooting ability.
CTRLsum	Yeovil Town Swansea City : Ander Herrera scored his side's opening goal against Aston Villa with his eyes shut. The 25-year-old has also scored without looking at either the ball or the opposition's net in games against QPR, Leicester City, Swansea and Yeovil Town.
	Manchester United Queens Park Rangers: Ander Herrera appeared to have his eyes closed as he scored against Aston Villa on Saturday. The 25-year-old has scored six of his seven goals without looking at the ball. Herrera also netted with his eyes open in Manchester United's 4-0 win over Queens Park Rangers.
	Aston Villa MUTV: Ander Herrera scored a brace against Aston Villa on Saturday. But the Spanish midfielder appeared to have his eyes closed . Louis van Gaal told MUTV that Herrera should be 'more composed' .
	Herrera Athletic Bilbao: Ander Herrera has scored six of his seven goals with his eyes closed. The former Athletic Bilbao man joined Manchester United for £29million in the summer.

Table 22: Random Paired Entity Control Examples

Article	A husband accused of trying to murder his cheating wife was ready to let her have sex with another man once a month as long as she stayed with him, his trial heard yesterday. A close friend told how Joseph O'Riordan, 74, had confided in him about the extraordinary plan for his 47-year-old wife Mandy. O'Riordan, a councillor and former nightclub owner, stabbed her eight times in a jealous rage after finding out she had been having an affair with a postman. Extraordinary deal: Joseph O'Riordan stabbed his wife of ten years Amanda (left) with a seven inch kitchen knife eight times - yesterday Brighton Crown Court heard he was considering allowing her to have affairs. She suffered life-threatening injuries after being knifed in the torso, chest, arms and back. The jury was also shown dramatic footage of the moment police arrived at the couple's home to be greeted by a 'calm' O'Riordan opening the door. The revelation of his proposal for keeping his wife of ten years came from Alfred Harris. He told how O'Riordan had confided five days before the attack that he believed she was having an affair. O'Riordan was 'choked up and emotional' when he said: 'I think Amanda is playing away. She's getting her nails and hair done more regularly, she's been on a diet and doesn't want sex.'. Asking for a suit: O'Riordan sent his wife this letter from his prison cell. The following day, added Mr Harris, the men met for a pub lunch in O'Riordan's home village of Polegate, East Sussex. 'I saw Joe and he told me that Amanda had been seeing someone else – a guy who drove a van. Joe said he loved Amanda to bits and if she wanted to have sex with someone else once a month that would be okay as long as she stayed with him'. In a statement read to Brighton Crown Court, Mr Harris also described the couple as 'loving and close' He was 'so shocked' to learn that O'Riordan had attacked his wife at their flat on a residential care home estate. The jury saw images of four police officers, one of whom was wearing a lapel camera, arriving shor
Reference Summary	Joseph O'Riordan, 73, stabbed wife eight times after discovering her affair. She was left with life-threatening injuries to her torso, chest, arms and back. Yesterday Brighton Crown Court heard about deal he was ready to offer her. He had told friend about the idea while in the pub just days before stabbing.
	Joseph O'Riordan Alfred Harris : Joseph O'Riordan, 74, is accused of stabbing wife Mandy, 47, eight times. Friend Alfred Harris told how he had told him about the extraordinary plan.
	Brighton Crown Court Stephen Drage : Joseph O'Riordan, 74, accused of stabbing wife Mandy, 47, eight times. Brighton Crown Court heard he was considering allowing her to have affairs. Dr Stephen Drage, an intensive care consultant, told jury how she was 'clearly hurt'
CTRLsum	Joseph O'Riordan Catt: Joseph O'Riordan, 74, is accused of stabbing wife Mandy, 47, eight times. PC Dave Catt said he 'lost it' when he found out about the affair.
	Stuart Kenway Joseph O'Riordan: Joseph O'Riordan, 74, is accused of stabbing wife Mandy, 47, eight times. PC Stuart Kenway told how he 'appeared calm and composed'.
	Brighton Crown Court Amanda: Joseph O'Riordan, 74, is accused of stabbing wife Amanda, 47, eight times. He denies attempted murder at Brighton Crown Court.

1198 I.3 Length Control

For length control, we randomly sample 3 test articles from CNNDM and for each article we generate five different-length summaries.

Table 23: Random Length Control Examples. Control tokens are bolded.

Article	Film director David Lynch has confirmed he will no longer direct the revival of "Twin Peaks" – a cult 1990s television show that was set to return in 2016. The offbeat TV series, created by Lynch and Mark Frost, featured a quirky FBI agent who went to the Pacific Northwest town of Twin Peaks to investigate the mysterious murder of a high school girl named Laura Palmer. The groundbreaking series is considered one of the most influential shows in television history. Lynch broke the news about his departure in a series of tweets, saying that the show's third season will continue without him. He said he felt the network was not offering enough money to produce the show "the way it needed to be done." Lynch also wrote that he had personally called the actors over the weekend to let them know he would no longer be directing. Showtime Network, which will air the nine-episode comeback, released a statement saying they were "saddened" by Lynch's decision. "We were saddened to read David Lynch's statement today since we believed we were working towards solutions with David and his reps on the few remaining deal points," read the statement. "Showtime also loves the world of Twin Peaks and we continue to hold out hope that we can bring it back in all its glory with both of its extraordinary creators, David Lynch and Mark Frost, at its helm." Showtime announced they would produce a third season in October last year. Actor Kyle MacLachlan, who played the coffee-obsessed FBI agent Dale Cooper in the original series, had confirmed he would reprise the lead role for the new season.
Reference Summary	David Lynch says he won't be directing new episodes of Twin Peaks. Showtime "saddened" over decision, which involved a dispute over money.
CTRLsum	 (Length bucket 0) director David Lynch longer direct revival Twin Peaks return 2016 season Showtime: Film director David Lynch will no longer direct the revival of "Twin Peaks" Showtime announced they would produce a third season in October last year. The show was set to return in 2016. (Length bucket 1) director David Lynch longer direct revival Twin Peaks set return 2016 season Showtime Network saddened: Film director David Lynch will no longer direct the revival of "Twin Peaks" Showtime Network say they are "saddened" by the decision. The show was set to return for a third season in 2016. (Length bucket 2) director David Lynch confirmed longer direct revival Twin Peaks set return 2016 series season continue Showtime Network saddened: Film director David Lynch has confirmed he will no longer direct the revival of "Twin Peaks" The series was set to return in 2016. Lynch broke the news about his departure in a series of tweets. He said the show's third season will continue without him. Showtime Network said they were "saddened" by Lynch's decision. (Length bucket 3) director David Lynch confirmed longer direct revival Twin Peaks 1990s set return 2016 series season continue Showtime Network saddened decision: SFilm director David Lynch has confirmed he will no longer direct the revival of "Twin Peaks" The legas "The 1990s TV series was set to return in 2016. Lynch broke the news about his departure in a series of tweets. The show's third season will continue without him. Showtime Network said they were "saddened" by Lynch's decision. (Length bucket 4) director David Lynch confirmed longer direct revival Twin Peaks cult 1990s set return 2016 series of tweets. The show's third season will continue without him. Showtime Network said they were "saddened" by Lynch's decision. (Length bucket 4) director David Lynch confirmed longer direct revival Twin Peaks cult 1990s set return 2016 series of tweets. The show's third season will continue wi

Table 24: Random Length Control Examples. Control tokens are bolded.

Article	Washington (CNN)An Iranian military observation aircraft flew within 50 yards of an armed U.S. Navy helicopter over the Persian Gulf this month, sparking concern that top Iranian commanders might not be in full control of local forces, CNN has learned. The incident, which has not been publicly disclosed, troubled U.S. military officials because the unsafe maneuver could have triggered a serious incident. It also surprised U.S. commanders because in recent months Iranian forces have conducted exercises and operations in the region in a professional manner, one U.S. and other world powers meet with Iran in Switzerland to negotiate a deal limiting Tehran's nuclear program. At the same time, Iran has been active in supporting proxies in several hotspots in the Persian Gulf and neighboring regions. The Navy MH-60R armed helicopter was flying from the deck of the USS Carl Vinson on a routine patrol in international airspace, the official said. An unarmed Iranian observation Y-12 aircraft approached. The Iranian aircraft made two passes at the helicopter, coming within 50 yards, before the helicopter moved off, according to the official. The official said the helicopter deliberately broke off and flew away in a 'predictable' manner so the Iranians could not misinterpret any U.S. intentions. The Navy helicopter was in radio contact with the ship during the encounter, but there was no contact between the two aircraft and no shots were fired. The Navy crew took photos of the incident but the military is not releasing them. The U.S. administration is considering a potential demarche protest against Iran, the official said. CNN has reached out to Iranian officials but has not received a response. This type of Iranian observation aircraft generally operates over the Gulf several times a month. But after the recent incident, U.S. naval intelligence did not see it again for two weeks, leading to the conclusion that the incident may have been ordered by a local commander who was then reprimanded by higher-ups. The Pentagon has
Reference Summary	Iranian plane came within 50 yards of U.S. Navy Sea Hawk copter. Navy copter was on patrol in international airspace. U.S. official think Iranian plane may have been under orders of local commander.
CTRLsum	 (Length bucket 0) Iranian military aircraft flew 50 yards armed U.S. Navy helicopter Gulf commanders: An Iranian military observation aircraft flew within 50 yards of an armed U.S. Navy helicopter over the Persian Gulf this month. The incident, which has not been publicly disclosed, has sparked concern that top Iranian commanders might not be in full control of local forces. (Length bucket 1) Iranian military observation aircraft flew 50 yards armed U.S. Navy helicopter Gulf Iranian commanders forces: An Iranian military observation aircraft flew within 50 yards of an armed U.S. Navy helicopter over the Persian Gulf. The incident sparked concern that top Iranian commanders might not be in full control of local forces.
	(Length bucket 2) Iranian military observation aircraft flew 50 yards armed U.S. Navy helicopter Persian Gulf Iranian commanders control forces incident : An Iranian military observation aircraft flew within 50 yards of an armed U.S. Navy helicopter over the Persian Gulf. The incident sparked concern that top Iranian commanders might not be in full control of local forces.
	(Length bucket 3) Iranian military observation aircraft flew 50 yards armed U.S. Navy helicopter Persian Gulf Iranian commanders control forces incident Iran protest: An Iranian military observation aircraft flew within 50 yards of an armed U.S. Navy helicopter over the Persian Gulf. The incident sparked concern that top Iranian commanders might not be in full control of local forces. Iran has been active in supporting proxies in several hotspots in the region. The U.s. administration is considering a potential demarche protest against Iran.
	(Length bucket 4) Iranian military observation aircraft flew 50 yards armed U.S. Navy helicopter Persian Gulf Iranian commanders control local forces incident officials official ordered Iran USS considering demarche protest Navy Aden: An Iranian military observation aircraft flew within 50 yards of an armed U.S. Navy helicopter over the Persian Gulf. The Navy MH-60R armed helicopter was flying from the deck of the USS Carl Vinson on a routine patrol in international airspace. The incident sparked concern that top Iranian commanders might not be in full control of local forces. "We think this might have been locally ordered," the official said. CNN has reached out to Iranian officials but has not received a response. Incident comes as the Navy patrols the Gulf of Aden to watch for Iranian ships. The administration is considering a potential demarche protest against Iran.

Table 25: Random Length Control Examples. Control tokens are bolded.

Article	It's a sight that draws giggles and curious stares from tourists and other first-timers – an unusual festival where revellers carry gigantic phalluses through the streets of a Japanese city. But for the residents of Kawasaki, who lug erotic shapes of all different sizes, this odd tradition is not a joke. Shinto Kanamara Matsuri started as a small tradition but has grown into a popular a tourist attraction, with participants praying to a god of fertility, child birth and protection from sexually transmitted infections. Participants carry a gigantic phallus through the streets of Kawasaki, Japan during the Shinto Kanamara Matsuri festival. The sight of three large phalluses being paraded through neighbourhoods in the city south of Tokyo draws giggles from tourists. Shinto Kanamara Matsuri, the Festival of the Steel Phallus, started as a small tradition but has grown into a popular a tourist attraction. Known as the Festival of the Steel Phallus, started as a small tradition but has grown into a popular a tourist attraction. Known as the Festival of the Steel Phallus, while spectators lick lollies or snack on sausages or vegetables shaped as male and female genitalia. Rainy weather didn't ruin the mood at this year's festival, which had a massive collection of foreigners, according to Japanese website RocketNews24. They watched as groups of locals carried three heavy phalluses modelled after a mikoshi portable shrine, which is commonly used in Shinto festivals. Residents of Kawasaki carry phalluses of all different sizes while participating in a tradition that began nearly 40 years ago. Participants pray to a god who is said to help with fertility, child birth and protection from sexually transmitted infections. Rainy weather didn't ruin the mood at this year's festival, which attracted thousands of spectators from Japan and abroad. Spectators can buy lollies, sausages and vegetables shaped as male and female genitalia while watching the unusual parade. The three mikoshi carried through the streets for hours are t
Reference Summary	Shinto Kanamara Matsuri started as a small tradition but has grown into a popular tourist attraction. Known as the Festival of the Steel Phallus, participants pray to a god of fertility and protection from infections. Rainy weather didn't ruin the mood at this year's festival, which attracted a large crowd of holidaymakers.
	(Length bucket 0) Shinto Kanamara Matsuri started tradition god fertility child birth protection infections Steel: Shinto Kanamara Matsuri, the Festival of the Steel Phallus, started as a small tradition. Participants pray to a god of fertility, child birth and protection from sexually transmitted infections.
	(Length bucket 1) Shinto Kanamara Matsuri started tradition praying god fertility child birth protection sexually infections Festival Steel: Shinto Kanamara Matsuri, the Festival of the Steel Phallus, started as a small tradition. Participants pray to a god of fertility, child birth and protection from sexually transmitted infections.
CTRLsum	(Length bucket 2) Shinto Kanamara Matsuri started tradition popular tourist praying god fertility child birth protection sexually infections Festival Steel Phallus: Shinto Kanamara Matsuri, the Festival of the Steel Phallus, started as a small tradition but has grown into a popular a tourist attraction. Participants pray to a god of fertility, child birth and protection from sexually transmitted infections.
	(Length bucket 3) Shinto Kanamara Matsuri started tradition popular tourist attraction praying god fertility child birth protection sexually transmitted infections Festival Steel Phallus: Shinto Kanamara Matsuri, the Festival of the Steel Phallus, started as a small tradition but has grown into a popular a tourist attraction. Participants pray to a god of fertility, child birth and protection from sexually transmitted infections.
	(Length bucket 4) phalluses Shinto Kanamara Matsuri started small tradition grown popular tourist attraction praying god fertility child birth protection sexually transmitted infections Kawasaki Japan Matsuri festival Tokyo Festival Steel Phallus 40 : Shinto Kanamara Matsuri started as a small tradition 40 years ago. It has grown into a popular tourist attraction in Kawasaki, Japan. The Festival of the Steel Phallus is held every spring at the Kanayama Shrine. Participants carry three phalluses through neighbourhoods south of Tokyo. They are praying to a god of fertility, child birth and protection from sexually transmitted infections. Thousands of tourists attended this year's festival, which raised money for HIV research.

I.4 Contribution Summarization on Scientific Papers

1201

Here we show three random examples from the arXiv test set. Note that this is the test set from (Cohan1202et al., 2018) instead of the contribution test data collected by us, because we want to show the difference1203between reference summaries (i.e. the paper abstract) in existing standard paper summarization dataset1204and our output contribution summaries. We truncate the source articles since they are too long to display.1205

Table 26: Random Contribution Summarization Examples. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

Article	synchronization of neural activity appears in different parts of the mammalian cerebral cortex @xcite, and underlies different neural processes in both normal and anomalous brain functions @xcite. it has been suggested that synchronization plays a vital role in information processing in the brain, e.g., processing information from different sensory systems to form a coherent and unified perception of the external world @xcite. on the other hand, synchronization has been detected in pathological conditions such as parkinson s disease @xcite . and epileptic seizures have long been considered resulting from excessive synchronized brain activity @xcite, though some recent studies suggest that this picture may be an over - simplification @xcite. therefore understanding the mechanisms of synchronization may be a critical step in elucidating how neural systems work @xcite . it has stimulated a great deal of theoretical and numerical works, such as the studies on the effects of the topological properties of underlying networks @ xcite and the dynamical properties of synaptic coupling @ xcite. it was recently shown that the response time of synaptic couplings influences the stability of synchronized oscillation in the nonlocally coupled hodgkin - huxley (hh) equations @ xcite. if the response time of synaptic coupling is slower, synchronization is favored by lower efficacy of synaptic transmission . the numerical studies @ xcite in a detailed computational model revealed that seizure - like activity occurs when the excitatory synapses are weakened , and the results were confirmed experimentally in mouse neocortical slices . according to the common accepted assumption that synchronization of neuronal activity underlies seizures , the dynamical mechanism of synchronization may be useful for understanding the way the biological neural system works. In this work, we numerically investigated the dynamical mechanism underlying the influence of synaptic efficacy on firing synchronization in the efficacy of synapse is low , n
Reference Summary	we investigated the influence of efficacy of synaptic interaction on firing synchronization in excitatory neuronal networks . we found spike death phenomena , namely , the state of neurons transits from limit cycle to fixed point or transient state . the phenomena occur under the perturbation of excitatory synaptic interaction that has a high efficacy . we showed that the decrease of synaptic current results in spike death through depressing the feedback of sodium ionic current . in the networks with spike death property the degree of synchronization is lower and unsensitive to the heterogeneity of neurons . the mechanism of the influence is that the transition of neuron state disrupts the adjustment of the rhythm of neuron oscillation and prevents further increase of firing synchronization .
CTRLsum	[the main contributions of this paper are : (1)]: we investigated the dynamical mechanism underlying the influence of synaptic efficacy on firing synchrony in hodgkin - huxley neuron networks; (2) we found that the dynamics of synaptic current plays an important role in determining the stability of firing synchronization.

Table 27: Random Contribution Summarization Examples. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

Article	for the understanding of surface reactions and the characterization of materials it is desirable to measure local forces close to a sample surface . the most common method to measure these surface forces is atomic force microscopy (afm)@xcite . historically , the first force measurements were static measurements for which the force is presented as a scalar function of the static tip - sample separation , the so - called force curve@xcite . this representation is sufficient for conservative forces but the total tip - surface force may also contain contributions from dissipative forces . since dissipative forces depend on probe velocity and past trajectory , dynamic force spectroscopy methods are required for their measurement . moreover , the visualization of dissipative forces as a function of position is valid only for a specific probe trajectory and simple force curves can not capture the full character of the interaction . despite the development of several dynamic methods@xcite surface forces are still usually treated as functions of the probe position only and represented by simple force curves . here , we present a comprehensive framework for the representation and analysis of complex surface forces as they are measured by dynamic afm . we concentrate on the most common modes of dynamic afm : amplitude - modulated afm (am - afm) and frequency - modulated afm (fm - afm) , which can be considered as narrow frequency band methods@xcite . we explore the fundamental limit of force reconstruction with marrow band dynamic afm at fixed probe height and show how minimal assumptions allow for a shere so of a cattlever is firmly clamped at one end and the tip is located at the other end which can move freqely . it is assumed that surface forces only act on the tip whereas the rest of the cantilever toos not experience significant surface force is upplied to maintain an oscillatory motion . thus , the dynamics are governed by the force between tip and surface , the extenal drive force and the properties of the ca
Reference Summary	in atomic force microscopy (afm) tip - surface interactions are usually considered as functions of the tip position only , so - called force curves . however , tip - surface interactions often depend on the tip velocity and the past tip trajectory . here , we introduce a compact and general description of these interactions appropriate to dynamic afm where the measurement of force is restricted to a narrow frequency band . we represent the tip - surface interaction in terms of a force disk in the phase space of position and velocity . determination of the amplitude dependence of tip - surface forces at a fixed static probe height allows for a comprehensive treatment of conservative and dissipative interactions . we illuminate the fundamental limitations of force reconstruction with narrow band dynamic afm and we show how the amplitude dependence of the p - surface interaction . with minimal assumptions this amplitude dependence force spectroscopy allows for a quantitative reconstruction of the effective conservative tip - surface force as well as a position - dependent damping factor . we demonstrate this reconstruction on simulated intermodulation afm data keywords _ : atomic force microscopy , measurement of force , mechanical resonators , mems / nems , dissipation , intermodulation
CTRLsum	[the main contributions of this paper are : (1)]: a comprehensive framework for the representation and analysis of complex surface forces as they are measured by dynamic atomic force microscopy (afm); (2) a study of the fundamental limit of force reconstruction with narrow band dynamic afm at fixed probe height and show how minimal assumptions allow for a quantitative reconstruction of the tip - surface interaction.

Table 28: Random Contribution Summarization Examples. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

Article	in this paper we discuss the mathematical aspects of the problems originating in the solution of nonlinear systems of hyperbolic partial differential equations . these equations describe a large variety of physical phenomena , such as , gasdynamics , magnetohydrodynamics (mhd) , shallow water equations , elasticity equations notivates the necessity of the development of reliable numerical methods for their solution . presence of discontinuous solutions motivates the necessity of the development of reliable numerical methods based on the fundamental mathematical properties of hyperbolic systems . although such methods are rather well developed for the euler gasdynamic equations in the conservation law form , their extension to more complicated hyperbolic systems is not straightforward . it requires a mathematical justification of the solution uniqueness , a formulation of the selection principles for relevant solutions , and , finally , an investigation of their physical validity . most of high resolution methods for gasdynamic equations use the exact or some of the approximate self - similar riemann problem solutions to determine fluxes through the computational cell surfaces . similar methods are expected to be developed for various types of hyperbolic systems . in this case we must construct the elementary self - similar solution using only admissible discontinuities (entropy consistent , evolutionary , etc.) . basically the choice of the solution must be made on the basis of the structure of the solution under vanishing viscosity and dispersion to create a proper background for the elecompetent of high - resolution numerical methods for hyperbolic systems more complicated than the euler equations of gasdynamics . we discuss several analytical and numerical solutions in the mentioned fields which illustrate the complexity of the selection problem and outline the methods of its solution . tvd upwind and symmetric differencing schemes have recently become very efficient tool for solving complex multi - shoc
	and comparison - this investigation requires 1) determination of a proper shope mining method in the parameter interpolation procedure necessary to obtain nonoscillatory schemes of the order of accuracy higher than one; ii) development of an efficient entropy correction method necessary to exclude rarefaction shocks; and , finally , iii) solution of the problem of excluding the origin of nonevolutionary solutions in ideal mhd calculations . the system of governing equations for a mhd flow of an ideal , infinitely conducting , perfect plasma in the cartesian coordinate system @xmath0 , @xmath1 , @xmath2 with the use of the conventional notations reads (one fluid approximation) : @xmath3 where @xmath4 is the vector of conservative variables and @xmath5 , @xmath6 , and @xmath7 are the flux vectors . we introduced here the source term @xmath8 in the form @xmath9 this form of the system can be used to satisfy the divergence - free condition by convecting away the magnetic charge from the computational region @xcite . otherwise , any other well - known method can be used to eliminate the magnetic charge . to determine a numerical flux @xmath10 normal to the computational cell boundary (@xmath11 is a unit outward vector normal to the cell surface) one can use the formulas based on the solution of the linearized problem @xmath12.] here @xmath13 and @xmath14 are the matrices formed by the right and by the left eigenvectors , respectively , of the frozen jacobian matrix @xmath15 the matrix @xmath16 is a diagonal matrix consisting of the frozen jacobian matrix eigenvalue moduli . the superscripts @xmath17 and @xmath18 denote the values at the right- and at the left - hand side of the cell boundary .
Reference Summary	a number of physical phenomena are described by nonlinear hyperbolic equations . presence of discontinuous solutions motivates the necessity of development of reliable numerical methods based on the fundamental mathematical properties of hyperbolic systems . construction of such methods for systems more complicated than the euler gas dynamic equations requires the investigation of existence and uniqueness of the self - similar solutions to be used in the development of discontinuity - capturing high - resolution numerical methods . this frequently necessitates the study of the behavior of discontinuities under vanishing viscosity and dispersion . we discuss these problems in the application to the magnetohydrodynamic equations , nonlinear waves in elastic media , and electromagnetic wave propagation in magnetics .
CTRLsum	[the main contributions of this paper are : (1)]: the mathematical aspects of the problems originating in the solution of nonlinear systems of hyperbolic partial differential equations; (2) the study of discontinuous solutions behavior under vanishing viscosity and dispersion to create a proper background for the development of high - resolution numerical methods for hyperbola systems more complicated than the euler equations of gasdynamics; and (3) solution of the problem of excluding the origin of nonevolutionary solutions in ideal magnetohydrodynamics calculations.

1206 I.5 Invention Purpose Summarization on Patent Filings

Here we show three random examples from the BIGPATENT test set. Note that this is the test set from
 origial BIGPATENT, because we want to show the difference between reference summaries in existing
 standard dataset and our output purpose summaries. We truncate the source articles since they are too
 long to display.

Table 29: Random Invention Purpose Summarization Examples. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

Article	referring to the drawings and , in particular to fig1 therein illustrated is a prior art surgical support mesh 10. mesh 10 may be manufactured from monofilament or multifilament yarns . prior art mesh 10, as shown , includes multifilament horizontally - extending yarns 12 and multifilament vertically - extending yarns 14 woven together to form a support trellis . the use of multifilament yarns , such as yarns 12 and 14 , provides a mesh having greater pliability and suppleness than the use of monofilament yarns . these characteristics result from both the smaller diameter of the individual filaments and the interstitial spaces or voids that are located between such filaments . in particular , the flexibility of a filament (art file b) generally increases as its diameter decreases . because the solid cross - sectional area of the filaments of a multifilament yarn is less than the cross - sectional area of a monofilament yarn of equivalent diameter , the multifilament yarn will have a greater degree of flexibility and pliability of filaments 16 that are intermingled or bundled together to form the yarn . interstitial spaces 18 , which are pockets of air , are formed between adjacent filaments of the yarn . although these voids contribute to the softness and pliability of the formed mesh , they also provide a natural breeding ground for bacteria or other infectious material . surgical mesh is , of course , thoroughly sterilized prior to implantation . nevertheless , surgeons typically prefer the use of monofilament - designed mesh to minimize any risk of infection . as a result , the advantages associated with multifilament - designed mesh (i. e. "softness and pliability which result in better assimilation of the mesh into the body) are typically sacrificed . It has been discovered therein that a surgical support mesh having both the softness and pliability of a multifilament vare wherein the interstitial voids between the filaments are enclosed within an infection - impervious matrix exhibits the desired resista
Reference Summary	a soft and pliable surgical support mesh exhibiting increased resistance to inhabitation of infectious matter . the mesh includes a support trellis formed of multifilament yarns wherein the interstitial voids located between the filaments of said yarns are enclosed within an infection - impervious matrix . the meshes may be designed to be extremely thin yet retain the requisite strength for repairing soft tissue , which allows for a low profile when folded for delivery .
CTRLsum	[the purpose of the present invention is]: to provide a surgical mesh that is resistant to the growth of bacteria and other infectious matter. this is accomplished by encapsulating the interstitial spaces located between the filaments of the yarn within a matrix.

Table 30: Random Invention Purpose Summarization Examples. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

Article	fig1 shows a multicolor web fed rotary printing press 1 in accordance with the invention . the press 1 includes four tower arrangements 2 a , 2 b , 2 c and 2 d for printing a single color or a multicolor image on the webs 4 a , 4 b , 4 c and 4 d . the webs 4 a , 4 b , 4 c and 4 d travel in a substantially linear direction through each of the towers 2 a - 2 d. For example , the web can travel along a substantially vorizontal direction , or in a substantially linear path at any desired angle relative to the vertical direction shown . the towers 2 a - 2 d each include four printing units 6 c , 6 m , 6 y and 6 b for respectively printing an image in (yan, magenta , yellow and black to no bch sides of each web 4 a - 4 d . other colors besides cyan, magenta , yellow and black to no bch sides of each web 4 a - 4 d . other colors besides cyan, magenta , yellow and black to no bch sides of each web 4 a - 4 d . other colors besides cyan, magenta , yellow and black to no bch sides of each web 4 a - 4 d . other colors besides cyan, magenta , yellow and black to no bch sides of each web 4 a - 4 d . other colors besides cyan, magenta , yellow and black to no bch sides of each web 4 a - 4 d . other colors besides cyan, magenta , yellow and black to no bch sides of each web 4 a - 4 d . other colors besides cyan, magenta , yellow and black to a be used . the webs 4 can be , for example , between 1200 and 1600 millimeters wide . each of the printing units 6 c , 6 m . 6 y and 6 b includes a ball screw 16 , which his rotatably supported in a housing 18 . the ball screw 16 can be rotated by a motor 20 as shown in fig2. fig2 shows one set of a spindle drive 10 , fixed b sindle 12 , ball screws 16 , and motors 20, but preferably each tower 2 is provided with several sets , one set for each corner of the print unit housing 18 . the motors 20 are controlled by a motor control unit 22 , which receives commands from a remote control 24 . by pressing a button on the remote control 24 , an operator can control the rotation of the motors 2
Reference Summary	the present invention is directed to a multicolor web fed rotary printing press having printing units that can be moved along a linear section of a web by a positioning mechanism to allow easy access to each of the printing units . the easy access to the printing units significantly reduces maintenance costs and downtime of the press . in addition , the invention provides a printing press that has a reduced overall size and allows the printing units to be arranged in a nested formation during printing operations . this nested formation reduces fanout and paper waste . the linear web section can be oriented vertically or horizontally .
CTRLsum	[the purpose of the present invention is]: to provide a web fed rotary printing press that allows an operator to access a desired part of a printing unit more easily than in the past.

Table 31: Random Invention Purpose Summarization Examples. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

Article	referring now to the drawings and more particularly to fig1 - 3, a fluid system 10 is illustrated and includes a variable displacement axial piston pump 12 that receives fluid from a tank 14 via a conduit 16 and delivers pressurized fluid via a supply conduit 18 to a fluid control valve 20 and selectively through work conduits 22, 24 to a fluid actuator 26. in the subject arrangement, the variable displacement axial piston pump 12 is a unidirectional pump that rotates in a counterclockwise direction as driven by a power input shaft 27. the fluid system 10 also includes first and second pressure sensors 28, 30 respectively connected to the tank conduit 16 and the supply conduit 18. the pressure sensors 28, 30 are operative to sense the pressure in the respective lines and deliver an electrical signal to a controller 32 through lectrical lines 34, 36. a position sensor 40 is mounted on the variable displacement axial piston pump 12 and operative to sense the displacement of the pump and deliver a signal representative thereof to the controller 32 via an electrical line 42. various other components could be used in the subject fluid system 10 without departing from the essence of the subject invention. For example, several control valves 20 and associated fluid attators 26 could be used . likewise , other sensors of various types and styles could be used . the variable displacement axial piston pump 12 includes a housing 44 having a head portion 46 and a body portion 48. the construction of the porting within the port plate 54 is disposed between the head portion 64 and the body portion 48. the construction of the porting illustrated in fig3 could be made within the head portion 64 ow thout departing from the essence of the subject invention . a rotating group 56 is disposed within the bed option 48 and includes a barrel 58 having a plurality of cylinder bores 59 is oriented within the barrel 58 parallel with the axis of rotation 60 of the barrel 58 having a plurality of cylinder bores 59 defined therein spaced fr
Reference Summary	a variable displacement axial piston pump is typically used to receive fluid from a tank and supply pressurized fluid through a control valve to move an actuator . the present variable displacement axial piston pump has a swashplate arrangement that is capable of being angled in two different directions to control the pressure transitions between the low pressure inlet port passage and the higher pressure outlet port passage as cylinder bores in a barrel of a rotating group rotate through trapped volume regions situated between inlet and outlet port passages of the axial piston pump . movement of the swashplate arrangement in two different directions provides smooth pressure transitions and increases the operating efficiency of the variable displacement axial piston pump .
CTRLsum	[the purpose of the present invention is]: to provide a variable displacement axial piston pump that is capable of delivering a variable amount of pressurized fluid in response to a change in the displacement of the pump.

I.6 Question-Guided Summarization

Table 32: Random Examples on Question-guided summarization. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

TEHRAN, Iran (CNN) - Iran's parliament speaker has criticized U.S. President-elect Barack Obama for saying that Iran's development of a nuclear weapon is unacceptable. Iranian President Mahmoud Ahmadinejad has outlined where he thinks U.S. policy needs to change. Ali Larijani said Saturday that Obama should apply his campaign message of change to U.S. dealings with Iran. "Obama must know that the change that he talks about is not simply a superficial changing of colors or tactics. Larijani said in comments carried by the semi-official Mehr News Agency. "What is expected is a change in strategy, not the repetition of objections to Iran's nuclear program, which will be taking a step in the wrong direction." In his first post-election news conference Friday afternoon, Obama reiterated that he believes a nuclear-armed Iran would be "unacceptable." He also said he would help mount an international effort to prevent it from happening. Larijani said that U.S. behavior toward Iran "will not change so simply" but that Obama's election showed internal conditions in the United States have shifted. He added that Iran does not mind if the United States provides other Persian Gulf countries with nuclear technology, but "you should know that you cannot prevent the Islamic Republic [from reaching its goals in the nuclear field]," according to the news agency. Obama cautioned Friday that it had only been a few days since the election and that he was not in office. "Obviously, how we approach Article and deal with a country like Iran is not something that we should simply do in a knee-jerk fashion. I think we've got to think it through," Obama said. "But I have to reiterate once again that we only have one president at a time. And I want to be very careful that we are sending the right signals to the world as a whole that I am not the president, and I won't be until January 20th." Larijani was speaking two days after Iranian President Mahmoud Ahmadinejad congratulated Obama, the first time an Iranian leader has offered such wishes to a U.S. president-elect since the 1979 Islamic Revolution. One analyst said the welcome was a gesture from the hard-line president that he is open to a more conciliatory relationship with the United States. Ahmadinejad said Tehran "welcomes basic and fair changes in U.S. policies and conducts," according to the state-run Islamic Republic News Agency on Thursday. Relations between the United States and Iran have historically been chilly and have been further strained in recent years over Iran's nuclear program. Tehran insists that the program exists for peaceful purposes, but the United States and other Western nations are concerned by Iran's refusal to halt uranium enrichment activities. CNN's Shirzad Bozorgmehr contributed to this report.

[Q: Iran criticizes who? A:]: U.S. President-elect Barack Obama. Gold Answer: U.S. President-elect Barack Obama

[Q: What are US and Iran relations tensioned about? A:]: Iran's nuclear program. Gold Answer: nuclear program.

CTRLsum [Q: Who said Obama should apply campaign message? A:]: Ali Larijani, Iran's parliament speaker. Gold Answer: Ali Larijani

[Q: What should Obama apply according to speaker? A:]: Change in strategy, not the repetition of objections. *Gold Answer: his campaign message of change*

[Q: What kind of weapons are being discussed? A:]: Nuclear weapons. *Gold Answer: nuclear*

Table 33: Random Examples on Question-guided summarization. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

Editor's note: The staff at CNN.com has recently been intrigued by the journalism of VICE, an independent media company and website based in Brooklyn, New York. VBS.TV is Vice's broadband television network. The reports, which are produced solely by VICE, reflect a transparent approach to journalism, where viewers are taken along on every step of the reporting process. We believe this unique reporting approach is worthy of sharing with our CNN.com readers. Brooklyn, New York (VBS.TV) - Whenever someone at our office bitches about being overworked, our stock response is "Beats digging ditches." While the express intention of the statement is usually not-so-supportive, we think it's a healthy reminder that at the end of the day, we are all basically professional e-mailers and should be thankful for such. The wildland firefighters who work for Grayback Forestry in Medford, Oregon, have no such motivational adages because their job is actually digging ditches. Around active forest fires. On the sides of mountains. You can't even bitch at these guys for having cushy government pensions to fall back on when they get older, because they're all private-sector contractors. Which means if they aren't out fighting forest fires or doing preventative forestry on unburned woods (basically extreme landscaping), they are losing money. They are the hardest working men in the tree business. Southern Oregon in the summer is a tinderbox. Last year the state recorded some 560 wild fires, the majority of which occurred in the seemingly endless sea of trees running across its bottom from the Cascades to the Pacific Coast. Humidity is next Article to nonexistent, which is extremely pleasant, but means that even an errant spark from a chainsaw or the proverbial cigarette butt out the car window can set the entire region ablaze. Flying into the Rogue Valley, there is evidence of past wild fires is everywhere: From the miles-long scar of the 2002 "Biscuit" fire stretching past the horizon, to the smaller pockets of charred trees crowding the edge of towns to the blue Wilderness-Firefighter-ribbon bumper stickers flying past on the I-5 to the elaborate wildfire and firefighter shrines at a local bar. See more of Oregon Fire Lines at VBS.TV There are very few places firemen aren't revered as local heroes, but the wilderness firefighters of Oregon go past people simply risking their lives to help others into a crazy superhero realm where their work regularly prevents entire settlements from being destroyed. They're literally the guardians of their communities. Incidentally, their work is also very beneficial for the forest. We spent a few days following a crew of Grayback forest-firefighters walk up the sides of what most people would consider a cliff, to chop down underbrush in preparation for a controlled burn. This is what they like to call "project work" - the light stuff they do between fires. The work is the hardest and least rewarding work we have ever tried to do. Unless you consider 12-hour-plus shifts of backbreaking labor, virtually zero outside recognition, and occasional accusations of being shills for the timber industry rewards. Which we do not. This piece was originally produced in August 2010.

[Q: What does the VBS call it? A:]: Project work. Gold Answer: "Beats digging ditches."

[Q: How long are the shifts? A:]: 12-hour shifts. *Gold Answer: 12-hour-plus*

[Q: What does the 12 hr shift involved? A:]: Project work.

CTRLsum Gold Answer: backbreaking labor, virtually zero outside recognition, and occasional accusations of being shills for the timber industry

[Q: Which group followed firefighters? A:]: The wildland firefighters of Oregon. *Gold Answer: VBS.TV*

[Q: What is the VBS following? A:]: Wildland firefighters who work around active forest fires. Gold Answer: a crew of Grayback forest-firefighters Table 34: Random Examples on Question-guided summarization. Control tokens are bolded. "[]" denote that the tokens are used as both keywords and prompts.

WASHINGTON (CNN) - The nation's largest publicly owned utility company may be vulnerable to cyber attacks, according to a new report. In 2007 President Bush visited the Browns Ferry Nuclear Plant, operated by the Tennessee Valley Authority. The Tennessee Valley Authority, which supplies power to almost 9 million Americans, "has not fully implemented appropriate security practices to protect the control systems used to operate its critical infrastructures," leaving them "vulnerable to disruption," the Government Accountability Office found. Simply put, that means a skilled hacker could disrupt the system and cause a blackout. Rep. James Langevin, a Rhode Island Democrat, fears the problem is much larger than just the TVA. "If they are not secure, I don't have a great deal of confidence that the rest of our critical infrastructure on the electric grid is secure," he said. The TVA operates 52 nuclear, hydroelectric and fossil-fuel facilities in the southeastern United States. Among the government watchdog agency findings: • The TVA's firewalls have been bypassed or are inadequately configured • Passwords are not effective • Servers and work stations lack key patches and effective virus protection • Intrusion-detection systems are not adequate • Some locations lack enough physical security around control systems. The GAO recommends 73 steps to correct the problems in its report to Congress. In September, CNN first aired dramatic footage of a government experiment demonstrating that a cyber attack could destroy electrical equipment. The experiment, dubbed "Aurora," caused a generator to fall apart and grind to a Article halt after a computer attack on its control system. The test was conducted by scientists at the Idaho National Laboratory. In October, the North American Electric Reliability Corp. told Congress that 75 percent of utilities had taken steps to mitigate the Aurora vulnerability, but Langevin said it now appears that Congress was misled. A congressional audit of the electric reliability corporation's claim cast doubt on the assertion that most utilities were taking steps to fix the problem. "It appears that they just made those numbers up," Langevin said. "It is not acceptable. It is outrageous." He said the result is there is now no clear picture of how vulnerable utilities are to cyber attacks. The electric reliability corporation - a nongovernmental group that oversees the power system and comprises members of the industry and some consumers - told CNN it regrets the confusion. Experts told CNN that Cooper Industries is the only manufacturer of hardware that can close the Aurora vulnerability. The company estimated it would need to sell about 10,000 devices to fix the problem nationwide. It has sold just over 100, it told CNN. Langevin said the federal government may need new powers to require utilities to take corrective actions to close cyber security gaps, and he will press to give those powers to the Federal Energy Regulatory Commission. The congressman is chairing an Emerging Threats, Cybersecurity, and Science and Technology subcommittee hearing Wednesday afternoon. Representatives of the TVA, the GAO, the federal commission and the electric reliability corporation are to appear before the subcommittee.

[Q: Are cyber attacks a real threat? A:]: Yes. Gold Answer: skilled hacker could disrupt the system and cause a blackout.

[Q: What steps are being taken to access vulnerabilities? A:]: The GAO recommends 73 steps. *Gold Answer: 73*

CTRLsum **Q: What percentage of utilities were fixed to combat attacks? A:**]: 75 percent. *Gold Answer: 75 percent*

> [Q: Who can disrupt the Tennessee Valley Authority? A:]: A skilled hacker. Gold Answer: skilled hacker

[Q: What was Congress told? A:]: Nothing. Gold Answer: 75 percent of utilities had taken steps to mitigate the Aurora vulnerability,