

FLEXCAD: UNIFIED AND VERSATILE CONTROL-LABLE CAD GENERATION WITH FINE-TUNED LARGE LANGUAGE MODELS

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ABSTRACT

Recently, there is a growing interest in creating computer-aided design (CAD) models based on user intent, known as controllable CAD generation. Existing work offers limited controllability and needs separate models for different types of control, reducing efficiency and practicality. To achieve controllable generation across all CAD construction hierarchies, such as sketch-extrusion, extrusion, sketch, face, loop and curve, we propose FlexCAD, a unified model by fine-tuning large language models (LLMs). First, to enhance comprehension by LLMs, we represent a CAD model as a structured text by abstracting each hierarchy as a sequence of text tokens. Second, to address various controllable generation tasks in a unified model, we introduce a hierarchy-aware masking strategy. Specifically, during training, we mask a hierarchy-aware field in the CAD text with a mask token. This field, composed of a sequence of tokens, can be set flexibly to represent various hierarchies. Subsequently, we ask LLMs to predict this masked field. During inference, the user intent is converted into a CAD text with a mask token replacing the part the user wants to modify, which is then fed into FlexCAD to generate new CAD models. Comprehensive experiments on public dataset demonstrate the effectiveness of FlexCAD in both generation quality and controllability.

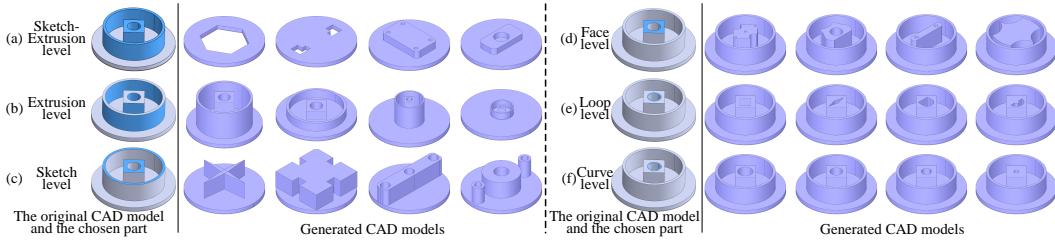


Figure 1: Controllable CAD generation achieved by FlexCAD. In each sub-figure, the left side shows the input: an original CAD model along with the part the user intends to modify (highlighted in blue). The right side displays the output: multiple new CAD models with only the chosen part changed. Users have the flexibility to specify the part in any CAD construction hierarchies, ranging from coarse levels like sketch-extrusion to fine levels like curve (as illustrated from (a) to (f)).

1 INTRODUCTION

A computer-aided design (CAD) model is a digital representation of a 2D or 3D object. It has been widely used across numerous industries, including architecture, product design and manufacturing, facilitating precise, efficient, and innovative development Ganin et al. (2021); Khan et al. (2024). In commonly used CAD tools like SolidWorks and AutoCAD, *sketch-and-extrude modeling* (SEM) is prevalent. This involves drawing 2D sketches and then extruding them into 3D shapes. Compared to other representations, such as Constructive Solid Geometry (CSG) Yu et al. (2024), B-rep Xu et al. (2024), or voxel Li et al. (2023) and point cloud Khan et al. (2024)-based formats, SEM, incorporating multiple CAD construction hierarchies including *sketch-extrusion*, *extrusion*, *sketch*,

054 *face*, *loop* and *curve* (see Fig. 3(a)), directly illustrates the drawing process of a 3D object. This
 055 allows for easy editing and reuse of CAD models, which is essential in CAD development.
 056

057 Recently, there is an increasing interest in developing generative models to automatically produce
 058 SEM of a CAD model¹. Specifically, DeepCAD Wu et al. (2021) focuses on uncontrollable gen-
 059 eration, where a CAD model is generated from a randomly sampled vector. However, providing
 060 controllability, *i.e.*, generating CAD models according to user intent, is crucial for the practical ap-
 061 plication of generative models. To address this, SkexGen Xu et al. (2022) and Hnc-cad Xu et al.
 062 (2023) implement disentangled codebooks to offer some levels of control. As each codebook en-
 063 codes a particular construction hierarchy, their controllability is quite restricted. For instance, Skex-
 064 Gen does not allow selecting a specific sketch for modifications when a CAD model comprises
 065 multiple sketches, nor can it handle finer-grained hierarchies such as faces and loops. Hnc-cad lacks
 066 control over the topology and geometry of curves. In summary, existing methods face challenges
 067 in providing adequate controllability across all CAD construction hierarchies. Additionally, they
 068 require separate models to deliver different types of control, which is inefficient and less practical.

069 The emergence of large language models (LLMs) offers insights for addressing these challenges.
 070 First, LLMs have exhibited remarkable success in handling diverse user queries with a single and
 071 unified model Chung et al. (2024). This phenomenon not only occurs in natural language tasks but
 072 also extends to other areas with domain-specific fine-tuning, such as human motion generation Jiang
 073 et al. (2024) and crystal material synthesis Gruver et al. (2024). Second, LLMs might have acquired
 074 CAD-related knowledge during the pre-training by learning CAD-specific codes, such as JSCAD
 075 codes Makatura et al. (2023). Third, prior to the rise of LLMs, small transformer-based models
 076 were explored for tasks like uncontrollable generation and image-to-sketch translation in the 2D
 077 sketch domain Ganin et al. (2021), showcasing the possibility of LLMs from a different perspective.

078 In this work, we introduce FlexCAD, a unified model designed for controllable CAD generation
 079 across all hierarchies by fine-tuning LLMs. As shown in Fig. 1, FlexCAD receives the original CAD
 080 model along with the part the user wants to modify (highlighted in blue). Here, users can specify the
 081 part in any hierarchy. FlexCAD then generates multiple new CAD models, altering only the selected
 082 part. To achieve these abilities, first, FlexCAD translates a CAD model into a *concise and structured*
 083 *text* (see Fig. 3). Specifically, in each sketch, the curve type (*e.g.*, a line) is directly represented as
 084 textual tokens. The numerical data indicating geometry (*e.g.*, point coordinates in a line) is converted
 085 into decimal integers and then into textual tokens. A special token is added to mark the end of each
 086 hierarchy. Tokens from the finer-level hierarchy are concatenated to form the representation for the
 087 coarser-level hierarchy. We use a similar way to convert each extrusion. Consequently, unlike the
 088 one-hot representation used in Xu et al. (2022), FlexCAD provides a concise text representation
 089 of a CAD model, facilitating easier processing and understanding by LLMs. Second, FlexCAD
 090 introduces a *hierarchy-aware masking* strategy to enable fine-tuning LLMs for various controllable
 091 CAD generation tasks (see Fig. 2). During training, we replace a hierarchy-aware field, which
 092 contains a sequence of tokens in the CAD text, with a mask token. This field can be set adaptably
 093 to reflect various hierarchies. Then, we ask LLMs to predict the masked field. To achieve this,
 094 we design prompt templates for all hierarchies, where the mask tokens are tailored to match the
 095 corresponding hierarchies. These templates are uniformly sampled at each epoch during the
 096 fine-tuning of LLMs. In this way, we ensure that the generation tasks for all hierarchies are learned in a
 097 single and unified model. Besides, unlike Xu et al. (2022; 2023) that requires multi-stage training,
 098 FlexCAD achieves end-to-end training. During inference, a CAD model is represented as a CAD text
 099 with a mask token replacing the part the user wants to change. The masked CAD text is fed into the
 100 fine-tuned LLMs to get predictions. After infilling the masked text with these predictions, FlexCAD
 101 produces CAD texts that can be rendered into new CAD models. Overall, our contributions are:
 102

- We propose FlexCAD, a unified and versatile model for controllable CAD generation across all hierarchies, including sketch-extrusion, extrusion, sketch, face, loop and curve.
- To the best of our knowledge, FlexCAD is the first to leverage LLMs for controllable CAD generation. It converts a CAD model into a brief, structured text and employs hierarchy-aware masking to fine-tune LLMs for various controllable CAD generation tasks.
- We conduct extensive experiments on public datasets. Despite its simplicity, FlexCAD greatly improves generation quality and controllability, showing its effectiveness on the tasks presented in this work and indicating potential for other CAD generation scenarios.

¹In the following, we will use CAD model to refer to SEM of a CAD model for brevity.

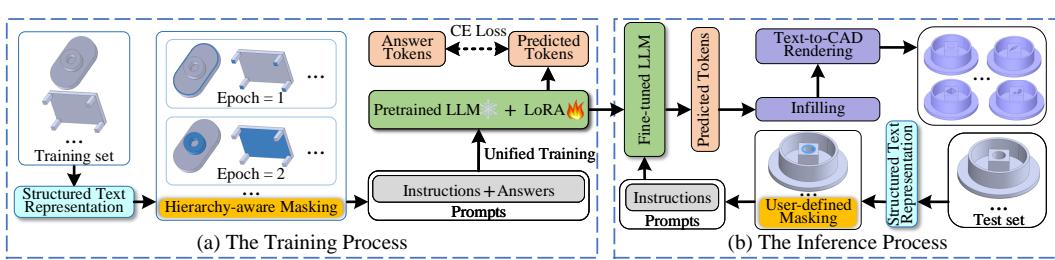
108 2 RELATED WORK 109

110 **CAD Model Generation.** A CAD model represents a 2D or 3D object digitally, with far-reaching
111 applications spanning multiple industries Li et al. (2024b). Existing CAD generation methods can
112 be classified into three categories based on representations of CAD models Khan et al. (2024): con-
113 structive solid geometry (CSG), boundary representation (B-rep) and sketch-and-extrude modeling
114 (SEM). CSG combines primitives (*e.g.*, cubes, cylinders, or spheres) via Boolean operations (*e.g.*,
115 union, subtraction or difference) to construct a CSG tree Laidlaw et al. (1986); Du et al. (2018);
116 Kania et al. (2020); Ren et al. (2021); Yu et al. (2022; 2024). B-rep characterizes a CAD model as a
117 graph Ansaldi et al. (1985), consisting of sets of interconnected faces, edges, and vertices Jayaraman
118 et al. (2023); Wang et al. (2022); Jayaraman et al. (2023); Xu et al. (2024).

119 Our work focuses on SEM, which directly models the drawing process of a CAD model, *i.e.*, drawing
120 2D curves to make sketches and then extruding them into 3D shapes. There are multiple construction
121 hierarchies in SEM, such as sketch-extrusion, extrusion, sketch, face, loop and curve. Compared to
122 CSG and B-rep, it enables easy editing and reuse of CAD models, which is critical in the CAD
123 design process. For SEM of a CAD model, DeepCAD Wu et al. (2021) utilizes a transformer-
124 based Vaswani et al. (2017) autoencoder for unconditional CAD generation. It merely designates
125 controllable generation as a future application. Nevertheless, offering controllability, *i.e.*, the ability
126 to generate CAD models based on user intents, is critical for enhancing the efficiency of CAD design
127 workflow in practical applications. SkexGen Xu et al. (2022) proposes to disentangle the topology
128 and geometry of sketches and extrusions to provide some levels of control. However, it does not
129 allow choosing a specific sketch, face, loop or extrusion for editing when multiple options exist.
130 Hnc-cad Xu et al. (2023) exploits three codebooks based on VQ-VAE framework Van Den Oord
131 et al. (2017) to control loop, sketch and extrusion level generation, respectively. While it offers finer
132 control, it struggles to manage the topology and geometry in the curve level. To sum up, despite
133 significant progress, existing work face challenges in offering controllability across all construction
134 hierarchies. Besides, they develop separate models to support different types of control.

135 **Large Language Models (LLMs).** LLMs have shown significant success recently Touvron et al.
136 (2023); Liu et al. (2024). Fine-tuning LLMs has demonstrated notable versatility and efficacy Luo
137 et al. (2024); Chung et al. (2024); Roziere et al. (2023); Ma et al. (2024); Zhang et al. (2024); Li et al.
138 (2024a). Specifically, the fine-tuned LLMs have displayed exceptional success by simultaneously
139 handling various generation tasks within a unified framework Wu et al. (2024); Jiang et al. (2024);
140 Gruver et al. (2024); Zou et al. (2024). Moreover, LLMs are generally pre-trained using extensive
141 high-quality datasets with cross-disciplinary and abundant knowledge Wenzek et al. (2020); Touvron
142 et al. (2023). For example, LLMs may have acquired CAD-related knowledge by learning CAD
143 codes, such as JSCAD codes Makatura et al. (2023). Encouraged by these achievements, we explore
the potential of using LLMs to tackle the aforementioned challenges in CAD generation.

144 3 METHODOLOGY 145



156 Figure 2: The overall framework of FlexCAD. (a) Training process. Initially, a CAD model is
157 converted into a structured text. Next, a hierarchy-aware masking strategy is proposed to mask a
158 specific field in the text with a special mask token. This field is set differently at each epoch to reflect
159 various hierarchies. Then, LLMs are fine-tuned to predict the masked field. (b) Inference process.
160 The original CAD model is transformed into a structured text with a mask token replacing the part
161 the user wants to change. The fine-tuned LLMs are provided with this masked text to generate
diverse predictions, which are then converted into new CAD models by infilling and rendering.

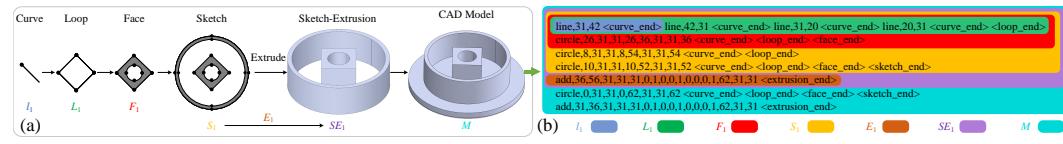


Figure 3: (a) An illustration for construction hierarchies of a CAD model. (b) Structured text representation for the CAD model shown in (a). The colors beneath the texts in (b) are used to indicate the relationship to construction hierarchies depicted in (a), e.g., blue for a curve and green for a loop.

In this section, we introduce FlexCAD, a unified model for controllable CAD generation across all construction hierarchies. As shown in Fig. 1, it receives an original CAD model along with the part the user wants to modify (highlighted in blue), and generates multiple new CAD models with only the selected part altered. To achieve this, as illustrated in Fig. 2, FlexCAD first translates a CAD model into a structured text (Sec. 3.1), and then introduces a hierarchy-aware masking strategy to enable fine-tuning LLMs for multiple controllable CAD generation tasks (Sec. 3.2)

3.1 REPRESENTING CAD AS STRUCTURED TEXT

Following the conventional definition of SEM of a CAD model Wu et al. (2021); Xu et al. (2022), there are multiple construction hierarchies, as illustrated in Fig. 3(a). A ‘curve’, i.e., line l , arc a , or circle c , forms the base level, represented by one, two, or four points respectively. Each point is denoted by its x and y coordinates. A ‘loop’ L denotes a closed path, comprising either a single curve (i.e., circle) or multiple curves (e.g., line-arc-line). A ‘face’ F is a 2D area, characterized by a single loop or an outer loop with one or multiple inner loops acting as holes. A ‘sketch’ S is composed of one or multiple faces, sharing a common extrusion command. An ‘extrusion’ E is a command that extends a sketch from a 2D plane into a 3D body. A ‘sketch-extrusion’ SE represents a single sketch-extrusion 3D body. A ‘CAD model’ M comprises one or multiple SE entities.

As shown in Fig. 3(b), we represent a CAD model as a succinct and structured text. Specifically, in each sketch, we start by representing a curve since it is the base level. The curve type (i.e., line, arc or circle) is represented directly as textual tokens. The point coordinates of the curve, which are numerical, are expressed as decimal integers and then converted into textual tokens. This contrasts with Xu et al. (2022) that uses binary representation for point coordinates. For example, when discretizing coordinates into a 64×64 grid, Xu et al. (2022) denotes the center coordinate as $([0, 1, 1, 1, 1, 1], [0, 1, 1, 1, 1, 1])$, while we represent it as $(31, 31)$. Next, the curve is denoted as a sequence of textual tokens, with the first one indicating its type and the others representing point coordinates (the text with a blue background in Fig. 3(b)). Notably, we add a special textual token ‘ H_end ’ to mark the end of each hierarchy, where $H \in \{\text{curve, loop, face, sketch, extrusion}\}$. This is also different from Xu et al. (2022) where one-hot vectors are used as ending flags. We concatenate tokens of multiple curves to create the representation for a loop (the text with a green background in Fig. 3(b)). Then, we use a similar way to form representations of other hierarchies, including face and sketch (the texts with red and yellow backgrounds in Fig. 3(b)). Furthermore, an extrusion can also be represented using textual tokens, with the first one specifying its type (e.g., add or cut) and the others denoting its numerical attributes (the text with a brown background in Fig. 3(b)). Finally, a complete CAD model is assembled by concatenating all the textual tokens from its sketch-extrusions (the text with a cyan background in Fig. 3(b), please see the meaning of numbers in the text in the appendix). Consequently, we convert a CAD model into a structured text, enabling efficient processing and comprehension by LLMs. Besides, our FlexCAD shortens the overall token length compared to Xu et al. (2022). Moreover, this text is straightforward to parse and interpret, thereby facilitating the implementation of the below hierarchy-aware masking strategy.

3.2 FINE-TUNING LLMs WITH HIERARCHY-AWARE MASK PREDICTION

In the following, with the structured text representation (referred to as CAD text for simplicity), we introduce how to fine-tune LLMs to develop a unified model for various controllable CAD generation tasks. In general, during training, a hierarchy-aware field in the CAD text is replaced with a mask token. The field, which consists of a sequence of tokens, can be designed to reflect different CAD construction hierarchies. Next, LLMs are asked to predict the masked field (see Fig. 2(a)).

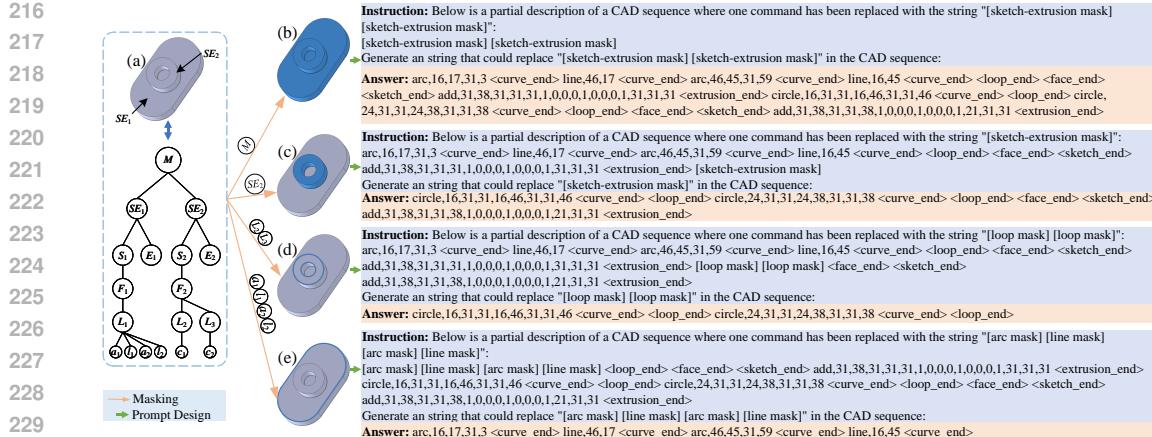


Figure 4: (a) illustrates a CAD model and its structural diagram. (b), (c), (d) and (e) are four examples for prompt templates with the mask tokens designed to represent different construction hierarchies. The masked field for different hierarchies in the CAD model are highlighted in blue.

To accomplish this, we design different prompt templates, where the mask tokens are designed to align with the corresponding hierarchies (see Fig. 4). During inference, given a CAD model, with a defined mask token, users can specify the part they want to modify (see Fig. 2(b)). Below, we further detail the design of prompt templates, the unified training and inference processes.

Prompt Template Design with Hierarchy-aware Mask. A prompt template includes an instruction with a special mask token replacing a hierarchy-aware field, and an answer containing the tokens of this field. Specifically, for the **CAD** level, we mask each internal sketch-extrusion with [sketch-extrusion mask]. In this case, other than the sketch-extrusion number, no information from the original CAD model is preserved. Fig. 4(b) shows an example. This allows us to freely generate CAD models with the expected number of sketch-extrusions during inference, facilitating the creation of CAD models with varying complexity. For the **sketch-extrusion**, **sketch** and **extrusion** levels, we replace the relevant field with [sketch-extrusion mask], [sketch mask], or [extrusion mask], respectively. Fig. 4(c) shows an example with one masked sketch-extrusion. For the **face (loop)** levels, given a face (loop), if it exclusively forms a sketch (face), we mask this face (loop) with [face mask] ([loop mask]). In cases where multiple faces (loops) belong to the same sketch (face), we use a corresponding number of mask tokens to mask them all at once. Fig. 4(d) illustrates an example, where two loops are replaced by two mask tokens. With this strategy, the model learns to generate faces (loops) with varying numbers as described in different instructions. For the **curve** level, all curves of the same loop are masked with their type indicated in the mask token (*i.e.*, line, arc or circle). Fig. 4(e) presents an example where four curves (arc-line-arc-line) belonging to the same loop are masked simultaneously. As the curve is the fundamental hierarchical level, the control of the topology and geometry of a sketch comes from it. Specifically, once trained and given a loop, by keeping its internal curve type and number unchanged, we can only modify the geometry. Alternatively, by varying the type or number of curves, we can alter the topology.

Unified Training by Sampling Prompt Templates. At each epoch, for a given CAD text, we uniformly sample a prompt template from the above seven hierarchies. The instruction in the template asks LLMs to predict the masked field autoregressively. Then, the cross-entropy (CE) loss between the prediction and the answer in the template is back-propagated to update the LLMs. To sum up, the advantage here is two-fold. First, by randomly choosing existing prompt templates at each epoch, we aim to establish a unified controllable generation model for various hierarchies. Second, beyond the existing prompt templates, we can incorporate new templates that support other tasks, such as unconditional generation. Notably, we fine-tune LLMs using LoRA Hu et al. (2022) which allows a few parameters to be trainable while keeping most parameters fixed. This allows us to leverage the advantages of large-scale models while accelerating model convergence Hu et al. (2022).

Inference with User-defined Mask. During inference, a CAD model is first converted to a CAD text, with a mask token replacing the part that needs modification. This masked CAD text is then

270 input into the fine-tuned LLMs to produce predictions. After infilling the masked text with these
 271 predictions, FlexCAD can provide various CAD texts that can be rendered into diverse CAD models.
 272 Notably, users do not have to strictly adhere to the masking pattern defined in the training process.
 273 For example, although the prompt template, used in training, masks all the loops tied to a face
 274 simultaneously, this is not mandatory in the inference. Due to the strong generalization capability
 275 of LLMs, it is possible to only mask a single loop for local editing, as illustrated in Fig. 2(b).

277 4 EXPERIMENTS

279 4.1 EXPERIMENTAL SETUP

281 **Datasets.** For consistency with prior work Xu et al. (2022; 2023), we evaluate our FlexCAD on
 282 the DeepCAD Wu et al. (2021) dataset. This dataset comprises 178,238 sketch-and-extrusion se-
 283 quences, divided randomly into training, validation, and test sets in a ratio of 90%-5%-5%. To
 284 ensure data quality, we follow SkexGen Xu et al. (2022) to remove duplicate and invalid sequences.
 285 Subsequently, we convert all resultant CAD sequences into texts, as mentioned in Sec. 3.1.

286 **Implementation Details.** We adopt the transformers Wolf et al. (2020) toolbox and select Llama-3-
 287 8B Meta (2024) as the base LLM, which achieves superior performance among open-source LLMs.
 288 For the 8B model, we use LoRA Hu et al. (2022) to fine-tune only 0.042% of their parameters,
 289 approximately 3.4 million. The LoRA rank and alpha are set to 8 and 32. The model is trained on
 290 four A6000 GPUs. we employ the AdamW optimizer Loshchilov & Hutter (2018), set the batch
 291 size to 32, use a cosine annealing learning rate of 5×10^{-4} , and train for 30 epochs. During the
 292 inference process, we set the sampling temperature τ and Top-p at 1.1 and 0.9, respectively.

293 **Metrics.** We adopt metrics consistent with previous methods Xu et al. (2022; 2023).
 294 Generally, the metrics Coverage (COV), Minimum Matching Distance (MMD) and
 295 Jensen-Shannon Divergence (JSD) measure generation diversity and quality on generated
 296 CAD models in comparison to the test set. See their detailed descriptions in Xu et al. (2023). Novel
 297 indicates the percentage of generated CAD models not present in the training set, while Unique
 298 represents the percentage of generated CAD models that appear only once within the generated set.
 299 Prediction Validity (PV) denotes the overall validity of predictions that can be rendered into 3D
 300 shapes rather than just 2D sketches or nothing. Realism denotes the realistic rate of generated CAD
 301 models compared to the training data, as assessed by human evaluators.

302 4.2 PERFORMANCE COMPARISION WITH EXISTING METHODS

304 **Baselines and Tasks.** We compare our FlexCAD with GPT-4o Achiam et al. (2023), one of the
 305 most powerful closed-source LLMs, and two state-of-the-art SEM-based baselines: SkexGen Xu
 306 et al. (2022) and Hnc-cad Xu et al. (2023). Since SkexGen and Hnc-cad cannot simultaneously
 307 control CAD generation across all hierarchies as ours, we choose the sketch-level and extrusion-
 308 level controllable generation tasks for comparison following the principles below. First, they are
 309 common tasks that can be handled by each baseline. Second, there are official implementations of
 310 baselines for these tasks. For GPT-4o, Hnc-cad and FlexCAD, given a CAD model from the test
 311 set, we randomly mask either a sketch or an extrusion and predict the corresponding masked field.
 312 Notably, despite our best efforts, SkexGen still has slightly different task settings compared to the
 313 above three methods. Specifically, when there are multiple sketches (or extrusions) in a CAD model,
 314 SkexGen changes all the sketches (or extrusions) instead of the specified one.

315 **Quantitative Comparison Results.** We randomly selected 1k CAD models from the test set. For
 316 every method, we generated 10 predictions for each model, resulting in a total of 10k CAD models.
 317 For the metrics COV, MMD, and JSD that require a subset of ground truths, we sampled 3k CAD
 318 models from the test set. The average scores across three runs are presented in Table 1. Without
 319 fine-tuning, GPT-4o performs poorly. On the other hand, our FlexCAD outperforms the baselines
 320 on nearly all evaluation metrics, demonstrating significant superiority in generation quality and
 321 controllability. Particularly, FlexCAD achieves the most notable improvement on PV, reaching up to
 322 20.8%-31.1% and 13.6%-44.5% in terms of sketch-level and extrusion-level controllable generation.

323 **Qualitative Comparison Results.** To illustrate the performance intuitively, we randomly selected
 three CAD models from the test set. As shown in Fig. 5, the results clearly illustrate that FlexCAD

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Table 1: Performance comparison on the DeepCAD test set. GPT-4o is enhanced with few-shot in-context learning. Specifically, each prompt comprises five exemplars randomly chosen from the training set. These exemplars include instructions and answers, following the format shown in Fig. 4. Moreover, we obtain the performance values of SkexGen and Hnc-cad based on their official codes. Best performances are in **bold**, and the second-bests are marked by *.

Method	Sketch-level						Extrusion-level							
	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑	Realism↑	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑	Realism↑
GPT-4o	58.2%	1.34	1.43	69.7%	72.8%	62.3%	23.2%	53.3%	1.42	2.14	58.6%	65.3%	48.8%	19.7%
SkexGen	60.6%	1.27	1.51	90.7%*	93.5%	68.7%	34.8%	63.6%	1.23	1.44	89.3%	89.1%*	76.1%	35.2%
Hnc-cad	62.4%*	1.21*	1.07*	87.6%	92.1%	72.6%*	36.3%*	65.6%*	1.25*	1.38*	86.2%	87.8%	79.7%*	38.0% *
Ours	65.6%	1.19	0.82	92.1%	92.6%*	93.4%	39.6%	68.5%	1.19	1.32	87.6%*	90.4%	93.3%	42.1%

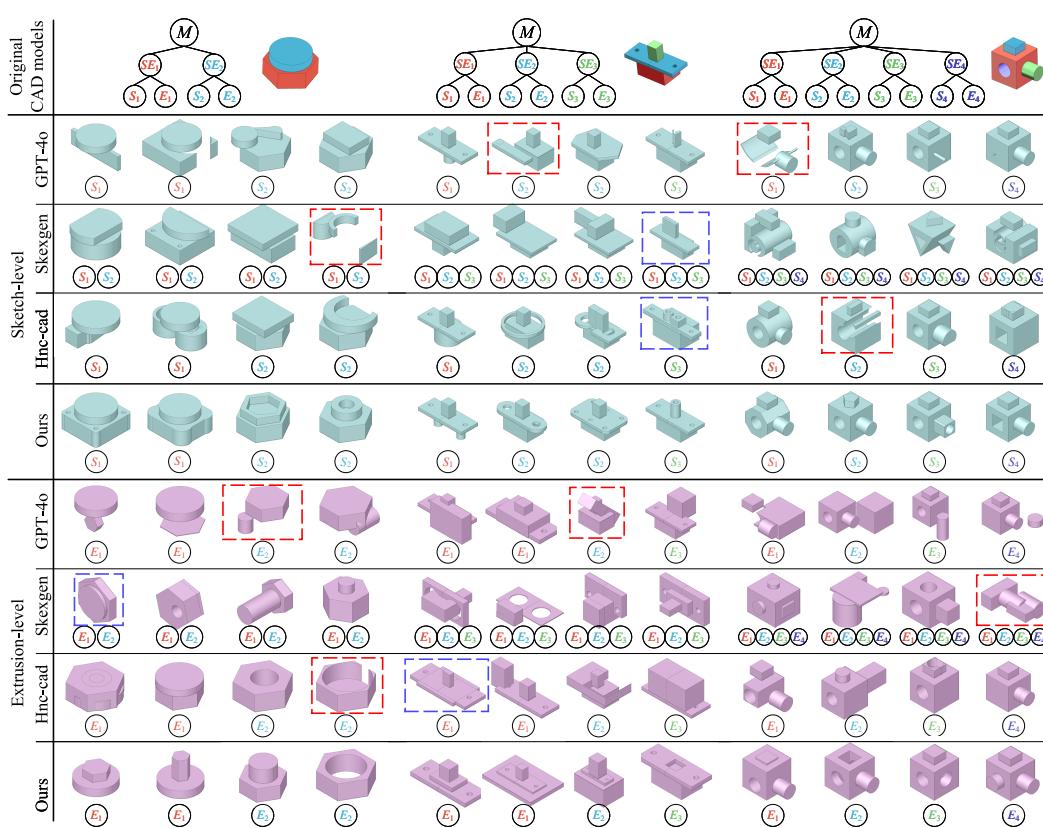


Figure 5: Qualitative comparison results for four methods. The first row displays three original CAD models, where the color of each sketch-extrusion aligns with that in the corresponding structural diagrams. In the following rows, given a CAD model, we randomly select its four newly predicted models for each method. The marks below the predictions are the corresponding masked and modified sketches or extrusions. The red boxes illustrate some of the most unrealistic examples. The blue boxes indicate some of the most obvious cases, where multiple fields simultaneously change in the same CAD model, rather than just the expected masked field.

greatly enhances the quality and controllability of CAD models compared to other competitors. Specifically, we tend to generate well-structured CAD models that closely resemble real-world examples, contrasting with unrealistic models like the red boxes shown in Fig. 5. On the other hand, as shown in the blue boxes in Fig. 5, SkexGen cannot specify which sketch or extrusion to modify, and Hnc-cad cannot preserve the integrity of the unmasked elements, even when recovering from the same codes. In contrast, we can mask any sketch or extrusion, ensuring that only the masked sketch or extrusion is modified while the remaining elements stay unchanged. This further confirms the effectiveness and superior controllability of our FlexCAD.

378	CAD level		Sketch-extrusion number = 1	Sketch-extrusion number = 2	Sketch-extrusion number = 3
379	Sketch-extrusion level				
380	Face level				
381	Loop level				
382	Curve level		4 lines	3 lines	3 lines + 1 arc
383	Original		Controllable Generation	Original	Controllable Generation
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385					
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391					

Figure 6: Our FlexCAD achieves controllable generation across different hierarchies, as introduced in 3.2. For the CAD level, we produce CAD models aligning with the required sketch-extrusion number. For the sketch-extrusion, face and loop levels, the left side of each sub-figure shows an original CAD model along with its local structural diagram. The color of each highlighted field matches that in the diagrams. The right side shows the predictions with only the masked part being masked and edited. And the masked part is marked below the predictions. Similarly, for the curve level, below the predictions are user-defined curve type and number. Best viewed in color.



Figure 7: Two examples of iterative editing. (a) Based on a simple CAD model, a new sketch-extrusion is generated by adding a [sketch-extrusion mask] at the end of the CAD text. Similarly, four peripheral internal loops are created. Then, a central quadrilateral loop is added and its geometry is altered. Finally, the new extrusion is adjusted to better match the original model. (b) Based on a complex CAD model, modifications can be progressively applied at various sketch-extrusions with the loop-extrusion-face-curve-loop level controllable generation to ultimately meet user needs.

Human Evaluation. To evaluate Realism, for each method, seven crowd workers were shown 950 pairs of images from the generated data and the training data, following Xu et al. (2023). They were asked to judge which of the two was more realistic. As shown in Table 1, for our FlexCAD, 39.6% and 42.1% of the generated models are more realistic. These rates are the highest among all methods, further demonstrating the advantage of our FlexCAD.

4.3 ENABLING MORE CONTROLLABLE GENERATION TASKS

In addition to the sketch and extrusion levels, our FlexCAD achieves controllable generation in other hierarchies, including the CAD, sketch-extrusion, face, loop, and curve levels. We provide some examples in Fig. 6, where users can modify the specific fields of a CAD model according to their intent. Additionally, we present detailed qualitative results across these hierarchies in Table 5 in the appendix. All the results together illustrate the effectiveness of our FlexCAD.

Furthermore, our FlexCAD can achieve iterative editing. For example, as shown in Fig. 7(a), starting from a simple CAD model, we iteratively edit it within a newly generated sketch-extrusion until the sketch-extrusion aligns closely with user requirements. On the other hand, as shown in Fig. 7(b), given a complex CAD model, we continuously adjust diverse hierarchy elements within different sketch-extrusions until they progressively match user specifications.

4.4 ABLATION STUDIES

We conduct several ablation studies evaluated on the sketch-level controllable generation as mentioned in Sec. 4.2, unless otherwise stated.

432
 433 Table 2: Ablation studies for fine-tuning LLMs with different settings. Pre-trained denotes
 434 the initial pre-trained weights. Full and LoRA indicate full and a few parameters are train-
 435 able, respectively. Transformer-4M is a small transformer-based Vaswani et al. (2017) language
 436 model. Its total number of trainable parameters is comparable to that of our model with LoRA.
 437 Llama-3-8B-From-Scratch and Llama-3-8B-Full denote training full parameters without and with
 438 the initial pre-trained weights, respectively. Llama-3-8B-Instruct is an instruction-tuned model in
 439 an 8B size Meta (2024). For Llama-3-70B Meta (2024), we fine-tune only 0.023% of its parame-
 440 ters, around 16.3 million. Best performances are in **bold** and the second-best are marked by *.

	Model	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑
441	Transformer-4M (w/o Pre-trained, Full)	59.4%	1.37	1.02	85.8%	86.9%	80.2%
442	Llama-3-8B-From-Scratch (w/o Pre-trained, Full)	63.0%	1.23	0.91	89.7%	90.2%	89.5%
443	Llama-3-8B-Full (w/ Pre-trained, Full)	66.4%*	1.20	0.85	92.6%*	91.0%	91.7%
444	Llama-3-8B-Instruct (w/ Pre-trained, LoRA)	65.3%	1.22	0.89	91.4%	92.1%*	90.5%
445	Llama-3-8B (w/ Pre-trained, LoRA, ours)	65.6%	1.19*	0.82	92.1%	92.6%	93.4%*
446	Llama-3-70B (w/ Pre-trained, LoRA)	68.2%	1.13	0.84*	93.0%	91.8%	94.6%

447
 448
 449 Table 3: Effectiveness analysis of the hierarchy-aware masking strategy and unified training.
 450 Random Masking denotes randomly masking 15%-50% continuous tokens within each CAD text,
 451 instead of the hierarchy-aware field. w/o Hierarchy-specific Tokens indicates that when masking,
 452 we utilize the generic token [mask], rather than employing hierarchy-specific mask tokens, such as
 453 [face mask], [loop mask] and etc. w/o Unified Training represents that we solely train a single
 454 task, i.e., the sketch-level controllable generation. Best performances are in **bold**.

	Model	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑
455	Random Masking	63.0%	1.25	1.02	88.2%	91.5%	90.6%
456	w/o Hierarchy-specific Tokens	63.7%	1.20	0.95	90.8%	91.7%	91.5%
457	w/o Unified Training	64.3%	1.17	0.89	91.6%	90.9%	92.2%
458	Ours	65.6%	1.19	0.82	92.1%	92.6%	93.4%

460
 461 **Settings of LLMs.** As shown in Table 2, without the pre-trained weights, both Transformer-4M
 462 and Llama-3-8B-From-Scratch achieve the lowest performances. This indicates that the pre-
 463 trained weights in LLMs contain valuable knowledge, which contributes to the performance gains.
 464 Llama-3-8B-Full achieves a performance similar to that of our model. However, it requires 80 hours
 465 to reach model convergence, compared to just 20 hours for ours. This highlights the effectiveness of
 466 the LoRA strategy Hu et al. (2022). The performance of Llama-3-8B-Instruct is slightly lower than
 467 that of our model. Conversely, as the model scale increases significantly, Llama-3-70B achieves the
 468 best performance but is more time-consuming and costly (See Sec. A.6 in the appendix).

469
 470 **Effectiveness of Hierarchy-Aware Masking.** As shown in Table 3, Random Masking exhibits the
 471 lowest performance, underscoring the effectiveness of masking the hierarchy-aware field. Further-
 472 more, the absence of hierarchy-specific tokens leads to a performance decline. In essence, these
 473 tokens potentially assist LLMs in more accurately determining the level they are generating, conse-
 474 quently enhancing overall controllable generation performance.

475
 476 **Effectiveness of Unified Training.** As depicted in Table 3, without unified training, there is a
 477 slight decrease in performance. In other words, when simultaneously training multiple controllable
 478 generation tasks across various levels, the inter-task knowledge contributes to performance gains.

479 5 CONCLUSION

480
 481 In this paper, we introduce a unified, versatile and user-friendly model, termed FlexCAD, which
 482 is particularly designed for controlling CAD generation across all hierarchies. To the best of our
 483 knowledge, we are the first to utilize LLMs for controllable CAD generation. Specifically, we
 484 convert each CAD model into a brief and structured text and propose hierarchy-aware masking for
 485 fine-tuning. Our FlexCAD is simple yet highly effective. Thorough qualitative and quantitative
 486 assessments conducted on public benchmarks confirm its effectiveness across all hierarchies.

486 **ETHICS STATEMENT**
 487

488 The data used in this work is tailored for creating and modifying CAD models. Due to its specialized
 489 nature, the misuse risk is naturally minimized, ensuring that the developed methods primarily benefit
 490 design and engineering tasks. In this work, we have invited crowd workers to evaluate the quality
 491 of generated CAD models. We conducted this work in accordance with ethical guidelines to ensure
 492 that participants were treated fairly, respectfully, and safely throughout the process. We took steps
 493 to protect the privacy of crowd workers by not collecting personally identifiable information. The
 494 data annotated by the crowd workers was used only for research purpose related to improving CAD
 495 generating techniques.

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648 649 Appendix

650 Considering the space limitation of the main paper, we provided more results and discussion in this
651 appendix, which is organized as follows:

- 653 • Section A: Additional Implementation Details and Analysis.
 - 654 – Sec. A.1: Meaning of Numbers in The CAD Text.
 - 655 – Sec. A.2: Detailed Results for Human Evaluation.
 - 656 – Sec. A.3: Quantitative results for Other Hierarchies.
 - 657 – Sec. A.4: Sensitivity Analysis of Key Hyper-parameters in Sampling.
 - 658 – Sec. A.5: Unconditional Generation Task.
 - 659 – Sec. A.6: Limitations and Future Work.
- 662 • Section B: Additional Qualitative Results.
 - 663 – Fig. 10: CAD Level Controllable Generation.
 - 664 – Fig. 11: Sketch-extrusion Level Controllable Generation.
 - 665 – Fig. 12: Extrusion level Controllable Generation.
 - 666 – Fig. 13: Sketch Level Controllable Generation.
 - 667 – Fig. 14: Face Level Controllable Generation.
 - 668 – Fig. 15: Loop Level Controllable Generation.
 - 669 – Fig. 16: Curve Level Controllable Generation.
 - 670 – Fig. 17: Unconditional Generation.

674 A ADDITIONAL IMPLEMENTATION DETAILS AND ANALYSIS

676 A.1 MEANING OF NUMBERS IN THE CAD TEXT.

679 Table 4: Effectiveness analysis of circle representation. To denote a circle, Center&Radius utilizes
680 the center coordinates along with the radius, while Diameter uses two uniformly distributed points
681 on the circumference that collectively define the diameter. Best performances are in **bold**.

682 Model	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑
683 Center&Radius	63.2%	1.21	0.87	90.2%	90.9%	89.7%
684 Diameter	66.4%	1.20	0.87	90.7%	91.5%	90.3%
685 Four points (Ours)	65.6%	1.19	0.82	92.1%	92.6%	93.4%

687 In this part, we explain the meaning of numbers in the CAD text, which builds upon SkexGen Xu
688 et al. (2022). In sketches, two (three) points can actually form a line (arc). In our work, a line, an
689 arc, and a circle are represented by one, two, and four points, respectively. Each point is denoted
690 as its x and y coordinates. Here, the second (third) point of a line (arc) is determined by the first
691 point of the subsequent curve (or the first curve when a loop is closed). Four points of a circle are
692 uniformly distributed along the circumference. We implement two variants of circle representation.
693 As shown in Table 4, when evaluated on the sketch-level controllable generation task, our FlexCAD
694 displays robustness across different circle representations, with Four points showing a slight edge.

695 Each extrusion operation is represented by 18 parameters: BVVTTTRRRRRRRRRSOO.

- 696 - B represents one of the three Boolean operations: add, cut or intersect. It occupies 1 parameter.
- 697 - V indicates the displacements of the top and the bottom planes from the reference plane in which
698 a sketch is extruded to form a solid. It occupies 2 parameters.
- 699 - T represents the 3D translation applied to the extruded solid. It occupies 3 parameters.
- 700 - R represents the 3D rotation of the extrusion direction. It occupies 9 parameters.
- 701 - S represents the uniform scaling factor. It occupies 1 parameter.
- O represents the center of scaling as a 2D coordinate. It occupies 2 parameters.

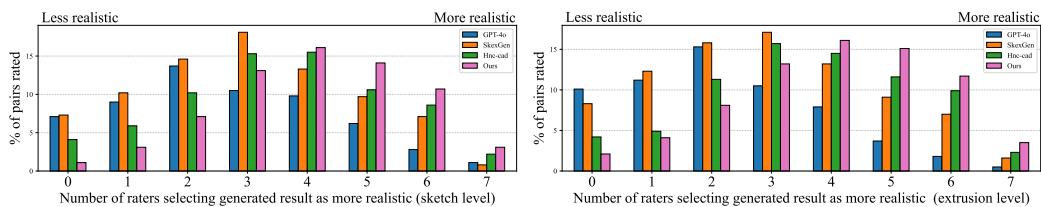


Figure 8: Distribution of realism scores from seven human evaluators. These scores are derived by comparing the generated CAD models produced by the four methods with the training samples.

A.2 DETAILED RESULTS FOR HUMAN EVALUATION

We present the detailed distribution of the Realism scores as mentioned in Table 1. As shown in the Fig. 8, the distributions for GPT-4o, SkexGen, and Hnc-cad are skewed towards the ‘less realistic’ end. Conversely, our FlexCAD demonstrates a primarily symmetric distribution, suggesting that the crowd workers struggle to differentiate between the generated models and the training set.

A.3 QUANTITATIVE RESULTS FOR OTHER HIERARCHIES

Table 5: Quantitative results for controllable generation across other hierarchies. The detailed evaluation setting is the same as that of Table 1.

Hierarchy	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑
CAD level	67.2%	1.14	0.77	92.6%	93.2%	91.8%
Sketch-extrusion level	65.3%	1.21	0.80	90.3%	89.7%	90.5%
Face level	62.9%	1.18	0.84	91.1%	90.9%	93.2%
Loop level	63.4%	1.15	0.81	88.3%	85.7%	90.5%
Curve level	59.1%	1.20	0.79	89.7%	91.5%	90.2%

In this section, we report the quantitative results across other hierarchies, including CAD, sketch-extrusion, face, loop and curve levels. By combining the data from Table 1 and Table 5, we observe that there is not a significant difference in performance across all hierarchies. These results together illustrate the effectiveness of our FlexCAD across all hierarchies.

A.4 SENSITIVITY ANALYSIS OF KEY HYPER-PARAMETERS IN SAMPLING

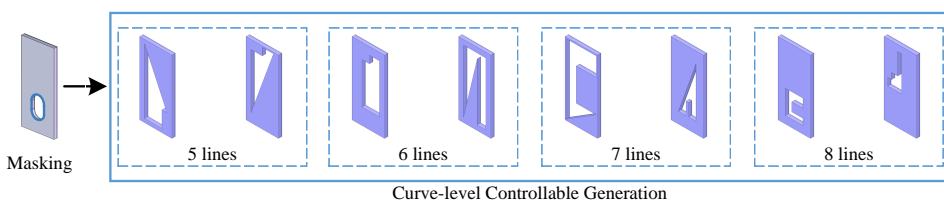
Table 6: Effectiveness analysis of key hyper-parameters, including the sampling temperature τ and Top-p. Best performances are in **bold** and the second-best are marked by *.

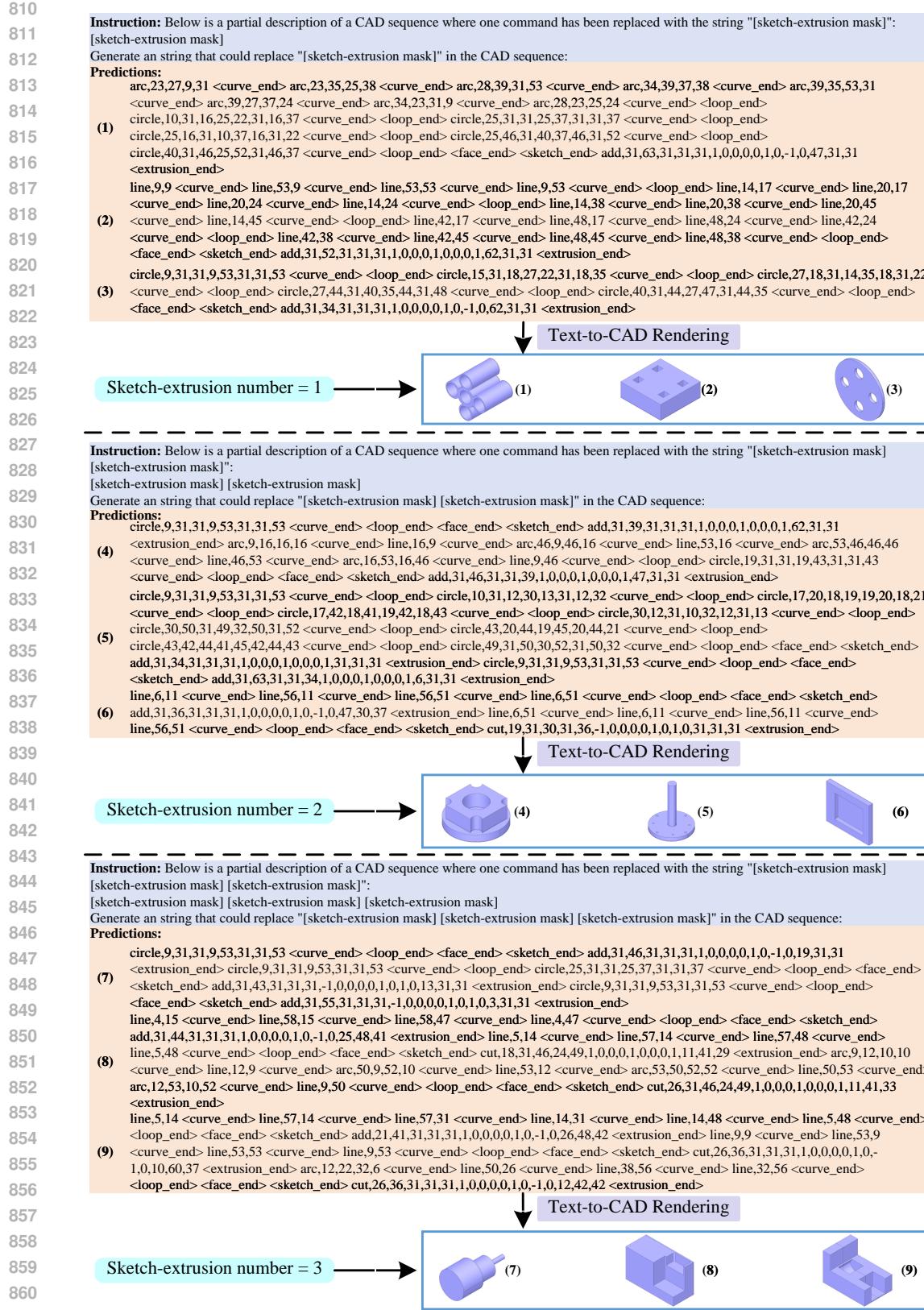
Model	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑
$\tau = 0.9$	62.7%	1.20	0.84	92.0%*	90.3%	96.7%
$\tau = 1.1$	65.6%*	1.19*	0.82*	92.1%	92.6%*	93.4%*
$\tau = 1.3$	65.8%	1.12	0.78	91.9%	94.5%	86.8%
Top-p = 0.8	62.9%	1.23	0.90	91.7%	87.8%	95.2%
Top-p = 0.9	65.6%*	1.19*	0.82*	92.1%*	92.6%*	93.4%*
Top-p = 1.0	68.3%	1.13	0.77	93.3%	95.7%	89.0%

In this part, we perform a sensitivity analysis on key hyperparameters in inference, including the sampling temperature τ and Top-p. As shown in Table 6, as either τ or Top-p increases, the performance of the first five metrics exhibits improvement, whereas the performance of the last deteriorates. Essentially, higher values of τ or Top-p lead to predictions that are more random and varied, while the overall prediction validity PV declines. In our experiments, we made a trade-off by selecting the values of τ and Top-p to guarantee that the PV value remains above 90%.

756 A.5 UNCONDITIONAL GENERATION TASK
757
758759 Table 7: Performance comparison for the unconditional generation task. Each method generates
760 10k CAD models, which are then compared with a randomly selected subset of 2.5k ground truth
761 models from the test set. For the baselines, other than the metric PV, we derive the values of other
762 metrics from the original paper Xu et al. (2023), while the values of PV are obtained based on their
763 official codes. Best performances are in **bold**, and the second-best are marked by *.
764

Model	COV↑	MMD↓	JSD↓	Novel↑	Unique↑	PV↑	Realism
SkexGen	84.7%	1.02	0.90*	99.1%*	99.8%	74.2%	46.9%
Hnc-cad	87.7%*	0.96*	0.68	93.9%	99.7%*	77.4%*	49.2%
Ours	89.2%	0.91	1.53	99.3%	96.9%	90.5%	51.5%

765 Our FlexCAD can easily achieve unconditional CAD generation by simply adding a prompt template
766 during training. Specifically, given a CAD text, the instruction in the prompt template can be as
767 concise as ‘*Below is a description of a CAD sequence:*’, while the corresponding answer is the
768 whole CAD text. The quantitative and qualitative results shown in Table 7 and Figure 17 verify
769 the effectiveness of our FlexCAD in unconditional CAD generation. Notably, as shown in Table 7,
770 our JSD exhibits the poorest performance. However, upon adjusting the sampling temperature τ or
771 Top-p to maintain the PV value at around 80%, the JSD value enhances significantly to 0.78.
772773 A.6 LIMITATIONS AND FUTURE WORK
774775 **Inference time.** We measure the inference time on one A6000 GPU (with a batch size of 1) for the
776 extrusion-level generation, averaging over 1,000 runs. The inference time of our FlexCAD (based on
777 Llama-3-8B) is slightly higher than that of SkexGen Xu et al. (2022) and Hnc-cad Xu et al. (2022),
778 at 0.56 seconds compared to 0.15 seconds and 0.38 seconds, respectively. Notably, we trained
779 and tested Llama-3-70B using four A100 GPUs (with a batch size of 1 per GPU), yet the average
780 inference time is still close to 3 seconds. Although LLMs demonstrate promising performance,
781 they generally lack efficiency. While the task of controllable CAD generation is not particularly
782 demanding in terms of real-time inference requirements, the slight increase in inference time of our
783 FlexCAD (based on Llama-3-8B) is acceptable given the promising performance.
784790 Figure 9: Failure cases. The generated CAD models display irregular polygonal shapes.
791792 **Failure cases.** Despite the significant progress, our FlexCAD sometimes generates hallucinations.
793 For example, when we implement curve-level controllable generation by replacing the original loops
794 with 5, 6, 7, or 8 lines, our FlexCAD tends to generate plausible CAD texts. The CAD models
795 produced from these texts typically exhibit irregular polygonal shapes, as illustrated in Fig. 9. In-
796 terestingly, when the modifications are limited to 4 lines, the generated predictions tend to be more
797 structurally regular, as illustrated in Fig. 16. We select loops composed of 4, 5, 6, 7, or 8 lines in the
798 training set for analysis. The ratio is approximately 77.9%-6.2%-9.5%-1.7%-4.7%. Consequently,
799 the fundamental reason behind these hallucinations may stem from data imbalance, which could
800 potentially be mitigated by collecting additional data.
801802 **Other limitations and future work.** In this paper, we fine-tune LLMs to achieve controllable CAD
803 generation. However, several unexplored domains warrant investigation. One such domain involves
804 determining how LLMs can be effectively utilized for text-based generation tasks, such as providing
805 a descriptive text like ‘*Create a desk.*’ to generate a corresponding CAD model. Since LLMs possess
806 robust capabilities in controllable CAD generation and text comprehension, in future work, we aim
807 to build a more advanced multi-modal LLM tailored for the text-based CAD generation task.
808



862 Figure 10: CAD level generation. Here, users can produce CAD models that match the required
863 sketch-extrusion number. The marks of predictions align with those of the images (the same below).

918
919

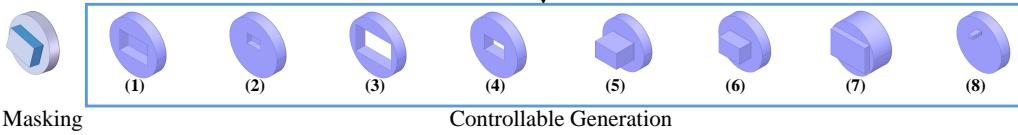
Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 circle,9,31,31,9,53,31,31,53 <curve_end> <loop_end> <face_end> <sketch_end> add,31,43,31,31,31,1,0,0,0,0,1,0,-1,0,62,31,31 <extrusion_end>
 line,5,14 <curve_end> line,57,14 <curve_end> line,5,48 <curve_end> <loop_end> <face_end> <sketch_end> [extrusion
 mask]

Generate an string that could replace "[extrusion mask]" in the CAD sequence:

Predictions:

- (1) cut,20,31,31,19,31,1,0,0,0,1,0,-1,0,31,31 <extrusion_end>
- (2) cut,25,31,31,19,31,1,0,0,0,1,0,-1,0,15,31,38 <extrusion_end>
- (3) cut,16,31,31,19,31,1,0,0,0,1,0,-1,0,39,31,31 <extrusion_end>
- (4) cut,18,31,31,19,31,1,0,0,0,1,0,-1,0,22,31,31 <extrusion_end>
- (5) add,31,57,31,19,31,1,0,0,0,1,0,-1,0,27,31,31 <extrusion_end>
- (6) add,31,47,31,19,31,1,0,0,0,1,0,-1,0,25,31,31 <extrusion_end>
- (7) add,31,43,31,19,31,1,0,0,0,1,0,-1,0,31,31,31 <extrusion_end>
- (8) add,31,43,31,19,31,1,0,0,0,1,0,-1,0,5,31,48 <extrusion_end>

Infilling & Text-to-CAD Rendering



Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 arc,9,12,10,10 <curve_end> line,12,9 <curve_end> arc,50,9,52,10 <curve_end> line,53,12 <curve_end> arc,53,50,52,52 <curve_end> line,50,53
 <curve_end> arc,12,53,10,52 <curve_end> line,9,50 <curve_end> <loop_end> <face_end> <sketch_end> circle,13,14,14,13,16,14,14,16 <curve_end> <loop_end>
 circle,13,48,14,46,16,48,14,49 <curve_end> <loop_end> circle,46,14,48,13,49,14,48,16 <curve_end> <loop_end> circle,46,48,48,46,49,48,48,49
 <curve_end> <loop_end> <face_end> <sketch_end> add,31,43,31,31,31,1,0,0,1,0,0,0,1,48,33,39 <extrusion_end> circle,9,31,31,9,53,31,31,53
 <curve_end> <loop_end> <face_end> <sketch_end> [extrusion mask]

Generate an string that could replace "[extrusion mask]" in the CAD sequence:

Predictions:

- (9) cut,22,31,33,38,43,1,0,0,0,1,0,0,1,25,31,31 <extrusion_end>
- (10) cut,25,31,33,38,43,1,0,0,0,1,0,0,1,13,28,31 <extrusion_end>
- (11) cut,0,31,33,38,43,1,0,0,0,1,0,0,1,31,31,31 <extrusion_end>
- (12) cut,22,31,32,38,43,1,0,0,0,1,0,0,1,6,31,31 <extrusion_end>
- (13) add,31,42,33,38,43,1,0,0,0,1,0,0,1,10,31,31 <extrusion_end>
- (14) add,31,46,33,38,43,1,0,0,0,1,0,0,1,22,29,31 <extrusion_end>
- (15) add,31,43,33,38,43,1,0,0,0,1,0,0,1,31,31,31 <extrusion_end>
- (16) add,31,41,33,38,43,1,0,0,0,1,0,0,1,5,31,31 <extrusion_end>

Infilling & Text-to-CAD Rendering

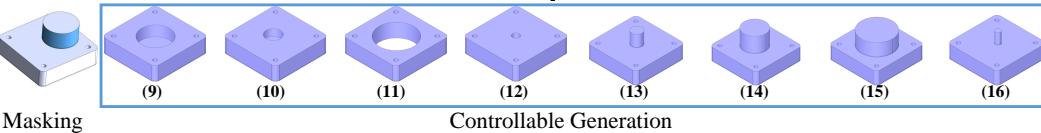


Figure 12: Extrusion level generation. Given a CAD model, users can mask any of its extrusions for modifications. The masked extrusions are highlighted in blue.

971

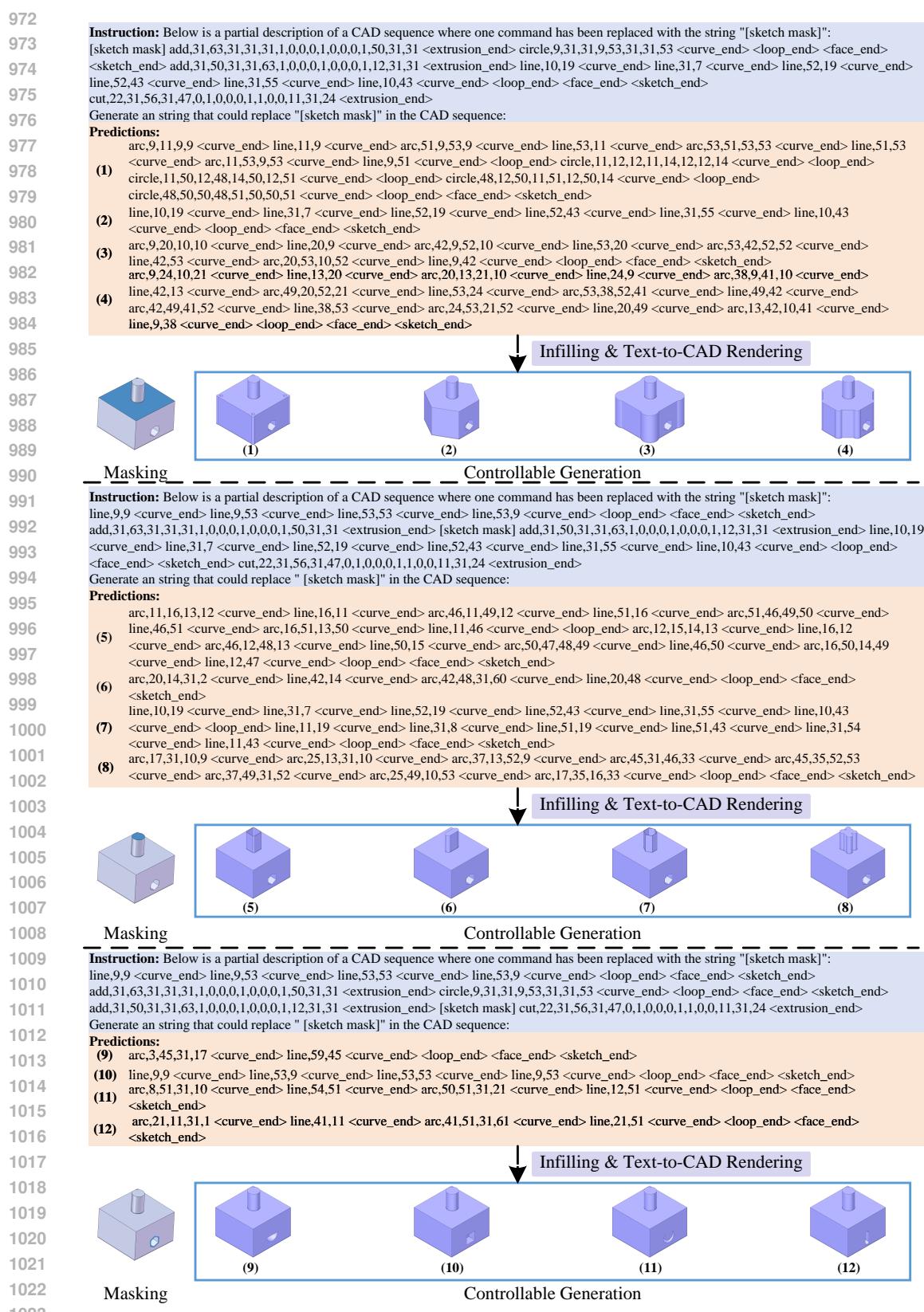


Figure 13: Sketch level generation. Given a CAD model, users can mask any of its sketches for modifications. The masked sketches are highlighted in blue.

Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[face mask] [face mask]":

```
line,11,7 <curve_end> line,51,7 <curve_end> line,51,55 <curve_end> line,11,55 <curve_end> <loop_end> <face_end> <sketch_end>
add,31,35,31,31,1,0,0,0,1,0,0,1,22,8,14 <extrusion_end> [face mask] [face mask] <sketch_end> add,31,41,10,16,35,1,0,0,0,1,0,0,0,1,9,30,33
<extrusion_end>
```

Generate an string that could replace "[face mask] [face mask]" in the CAD sequence:

Predictions:

- (1) line,0,26 <curve_end> line,16,26 <curve_end> line,16,36 <curve_end> line,0,36 <curve_end> <loop_end> <face_end> line,46,26 <curve_end>
 line,62,26 <curve_end> line,62,36 <curve_end> line,46,36 <curve_end> <loop_end> <face_end>
- (2) circle,0,31,6,25,12,31,6,37 <curve_end> <loop_end> <face_end> circle,50,31,56,25,62,31,56,37 <curve_end> <loop_end> <face_end>
- (3) circle,23,5,29,0,34,5,29,11 <curve_end> <loop_end> <face_end> circle,23,57,29,51,34,57,29,62 <curve_end> <loop_end> <face_end>
- (4) line,17,3 <curve_end> line,45,3 <curve_end> line,45,17 <curve_end> line,17,17 <curve_end> <loop_end> <face_end> line,17,45 <curve_end>
 line,45,45 <curve_end> line,45,59 <curve_end> line,17,59 <curve_end> <loop_end> <face_end>

↓ Infilling & Text-to-CAD Rendering

Masking **Controllable Generation**

Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[face mask]":

```
line,11,7 <curve_end> line,51,7 <curve_end> line,51,55 <curve_end> line,11,55 <curve_end> <loop_end> <face_end> <sketch_end>
add,31,35,31,31,1,0,0,0,1,0,0,1,22,8,14 <extrusion_end> line,7,11 <curve_end> line,17,11 <curve_end> line,17,51 <curve_end> line,7,51
<curve_end> <loop_end> <face_end> [face mask] <sketch_end> add,31,41,10,16,35,1,0,0,0,1,0,0,0,1,9,30,33 <extrusion_end>
```

Generate an string that could replace "[face mask]" in the CAD sequence:

Predictions:

- (5) circle,35,31,45,21,55,31,45,41 <curve_end> <loop_end> circle,41,31,45,27,49,31,45,35 <curve_end> <loop_end> <face_end>
- (6) circle,42,31,51,22,60,31,51,40 <curve_end> <loop_end> <face_end>
- (7) arc,26,9,31,3 <curve_end> line,36,9 <curve_end> arc,36,53,31,59 <curve_end> line,26,53 <curve_end> <loop_end> <face_end>
- (8) arc,45,11,55,11 <curve_end> line,45,51 <curve_end> <loop_end> <face_end>

↓ Infilling & Text-to-CAD Rendering

Masking **Controllable Generation**

Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[face mask]":

```
line,11,7 <curve_end> line,51,7 <curve_end> line,51,55 <curve_end> line,11,55 <curve_end> <loop_end> <face_end> <sketch_end>
add,31,35,31,31,1,0,0,0,1,0,0,0,1,22,8,14 <extrusion_end> [face mask] line,43,13 <curve_end> line,55,13 <curve_end> line,55,51 <curve_end>
line,43,51 <curve_end> <loop_end> <face_end> <sketch_end> add,31,41,10,16,35,1,0,0,0,1,0,0,0,1,9,30,33 <extrusion_end>
```

Generate an string that could replace "[face mask]" in the CAD sequence:

Predictions:

- (9) line,7,11 <curve_end> line,16,11 <curve_end> line,16,27 <curve_end> line,7,27 <curve_end> <loop_end> <face_end>
- (10) line,7,11 <curve_end> line,35,11 <curve_end> line,35,51 <curve_end> line,7,51 <curve_end> <loop_end> <face_end>
- (11) circle,1,31,11,21,21,31,11,41 <curve_end> <loop_end> <face_end>
- (12) arc,13,7,15,5 <curve_end> line,17,7 <curve_end> line,17,31 <curve_end> arc,17,55,15,57 <curve_end> line,13,55 <curve_end> <loop_end>
 <face_end>

↓ Infilling & Text-to-CAD Rendering

Masking **Controllable Generation**

Figure 14: Face level generation. Given a sketch, users can mask one or multiple faces within it for modifications. The masked faces are highlighted in blue.

Figure 14: Face level generation. Given a sketch, users can mask one or multiple faces within it for modifications. The masked faces are highlighted in blue.

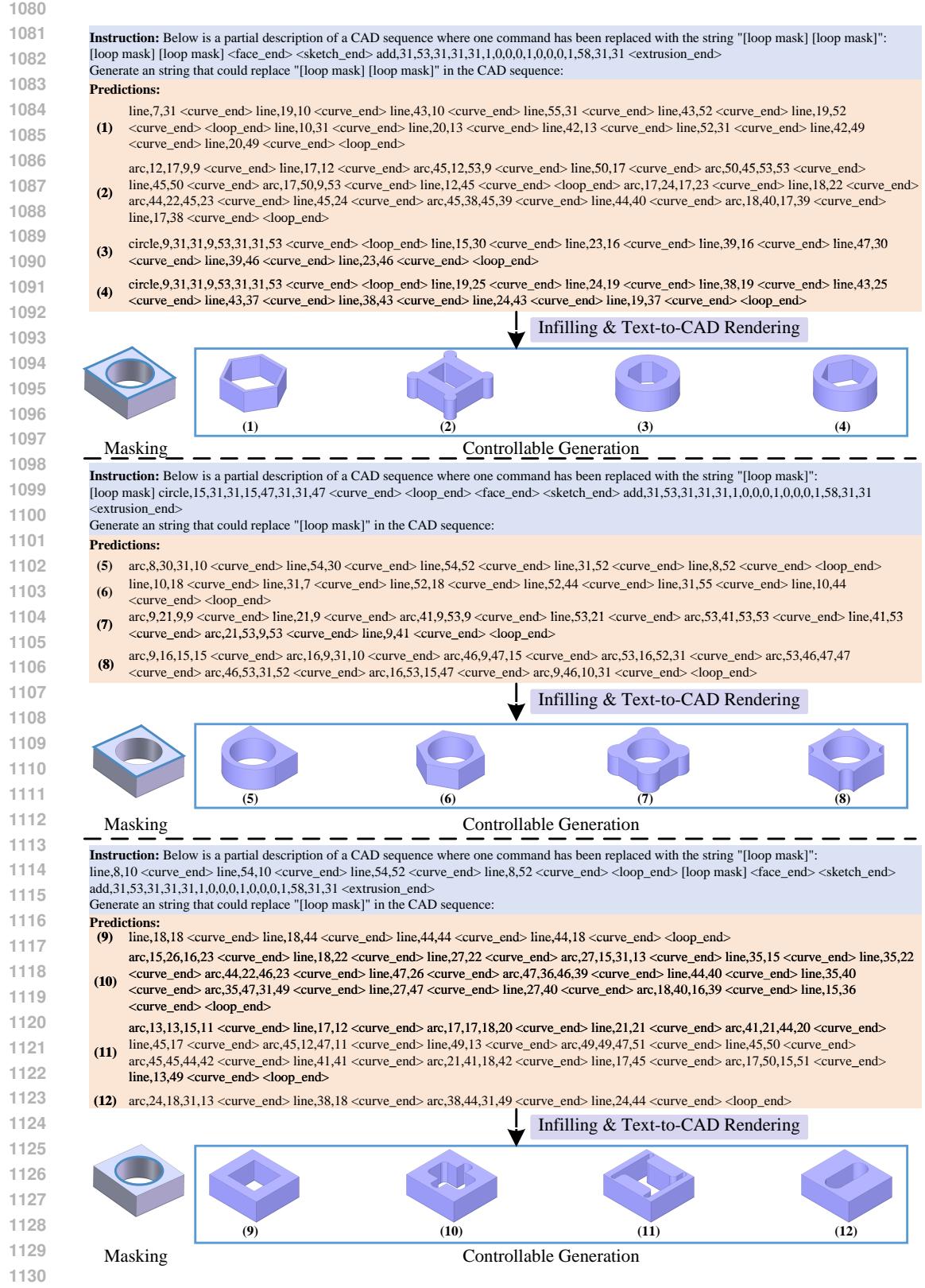
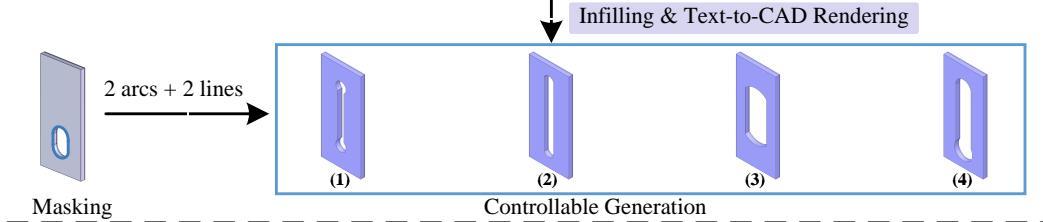


Figure 15: Loop level generation. Given a face, users can mask one or multiple loops within it for modifications. The masked loops are highlighted in blue.

Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[arc mask] [line mask] [arc mask] [line mask]":
line,18,2 <curve_end> line,44,2 <curve_end> line,44,60 <curve_end> line,18,60 <curve_end> <loop_end> [arc mask] [line mask] [arc mask] [line mask] <loop_end> <face_end> <sketch_end> add,31,34,31,31,31,1,0,0,0,0,1,0,-1,0,48,31,31 <extrusion_end>
Generate a string that could replace "[arc mask] [line mask] [arc mask] [line mask]" in the CAD sequence:
Predictions:
(1) arc,29,17,31,11 <curve_end> line,33,17 <curve_end> arc,33,45,31,51 <curve_end> line,29,45 <curve_end>
(2) arc,28,10,31,8 <curve_end> line,34,10 <curve_end> arc,34,52,31,54 <curve_end> line,28,52 <curve_end>
(3) arc,24,18,31,15 <curve_end> line,38,18 <curve_end> arc,38,44,31,47 <curve_end> line,24,44 <curve_end>
(4) arc,26,15,31,6 <curve_end> line,36,15 <curve_end> arc,36,47,31,52 <curve_end> line,26,47 <curve_end>



Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[arc mask] [line mask] [arc mask] [line mask] [arc mask] [line mask] [arc mask] [line mask]".

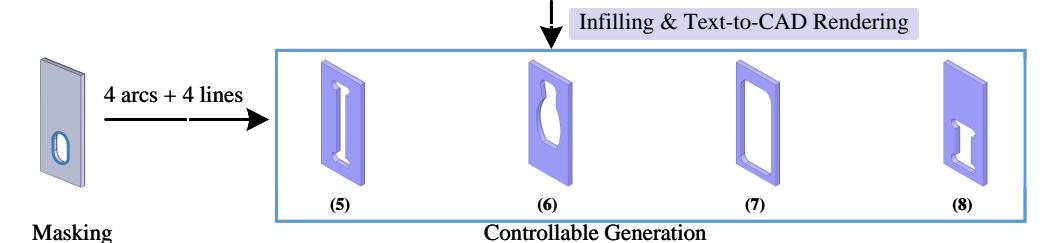
line,18,2 <curve_end> line,44,2 <curve_end> line,44,60 <curve_end> <loop_end> [arc mask] [line mask] [arc mask] [line mask] [arc mask] [line mask] [arc mask] [line mask]

<extrusion_end>

Generate an string that could replace "[arc mask] [line mask] [arc mask] [line mask] [arc mask] [line mask] [arc mask] [line mask]" in the CAD sequence:

Predictions:

- (5) arc,28,13,25,11 <curve_end> line,28,8 <curve_end> arc,34,8,37,11 <curve_end> line,34,13 <curve_end> arc,34,49,37,51 <curve_end> line,34,54 <curve_end> arc,28,54,25,51 <curve_end> line,28,49 <curve_end>
- (6) arc,27,43,23,31 <curve_end> line,27,19 <curve_end> arc,36,19,39,31 <curve_end> line,36,43 <curve_end> arc,36,50,36,53 <curve_end> line,33,55 <curve_end> arc,29,55,26,53 <curve_end> line,26,50 <curve_end>
- (7) arc,21,8,22,6 <curve_end> line,24,5 <curve_end> arc,38,5,40,6 <curve_end> line,41,8 <curve_end> arc,41,47,40,51 <curve_end> line,38,53 <curve_end> arc,24,53,22,51 <curve_end> line,21,47 <curve_end>
- (8) arc,26,11,23,9 <curve_end> line,26,6 <curve_end> arc,36,6,39,9 <curve_end> line,36,11 <curve_end> arc,36,26,39,28 <curve_end> line,36,31 <curve_end> arc,26,31,23,28 <curve_end> line,26,26 <curve_end>



Instruction: Below is a partial description of a CAD sequence where one command has been replaced with the string "[line mask] [line mask] [line mask] [line mask]".
line,18,2 <curve_end> line,44,2 <curve_end> line,44,60 <curve_end> line,18,60 <curve_end> <loop_end> [line mask] [line mask] [line mask] [line mask] <loop_end> <face_end> <sketch_end> add,31,34,31,31,31,1,0,0,0,0,1,0,-1,0,48,31,31 <extrusion_end>
Generate an string that could replace "[line mask] [line mask] [line mask] [line mask]" in the CAD sequence:
Predictions:
(9) line,23,7 <curve_end> line,39,7 <curve_end> line,39,55 <curve_end> line,23,55 <curve_end>
(10) line,26,20 <curve_end> line,36,20 <curve_end> line,36,42 <curve_end> line,26,42 <curve_end>
(11) line,24,7 <curve_end> line,38,7 <curve_end> line,38,31 <curve_end> line,24,31 <curve_end>
(12) line,19,3 <curve_end> line,43,3 <curve_end> line,43,59 <curve_end> line,19,59 <curve_end>

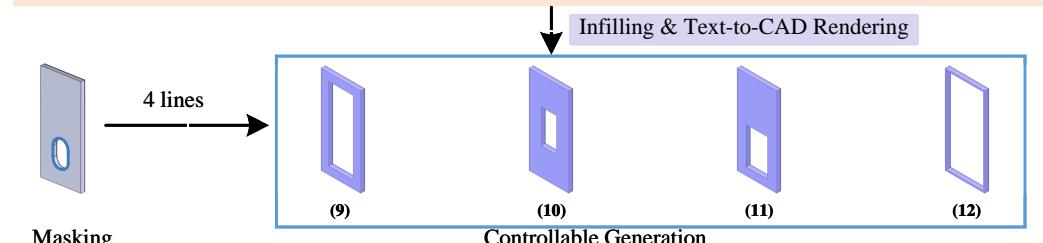


Figure 16: Curve level generation. Given a loop, users can mask all curves within it and specify the curve type and number of newly generated loops. The masked curves are highlighted in blue

Instruction: Below is a description of a CAD sequence:

Predictions:

```

arc,9,26,14,14 <curve_end> arc,26,9,31,11 <curve_end> arc,36,9,48,14 <curve_end> arc,53,26,51,31 <curve_end> arc,53,36,48,48
<curve_end> arc,36,53,31,51 <curve_end> arc,26,53,14,48 <curve_end> arc,9,36,11,31 <curve_end> <loop_end>
circle,11,25,16,20,22,25,16,31 <curve_end> <loop_end> circle,11,37,16,31,22,37,16,42 <curve_end> <loop_end>
circle,18,16,23,11,28,16,23,21 <curve_end> <loop_end> circle,18,46,23,41,28,46,23,51 <curve_end> <loop_end>
circle,27,31,31,27,35,31,31,35 <curve_end> <loop_end> circle,34,16,39,11,44,16,39,21 <curve_end> <loop_end>
circle,34,46,39,41,44,46,39,51 <curve_end> <loop_end> circle,40,25,46,20,51,25,46,31 <curve_end> <loop_end>
circle,40,37,46,31,51,37,46,42 <curve_end> <loop_end> <face_end> <sketch_end> add,31,37,31,31,31,1,0,0,0,0,1,0,-1,0,62,31,31
<extrusion_end>
line,5,13 <curve_end> line,57,13 <curve_end> line,57,49 <curve_end> line,5,49 <curve_end> <loop_end> circle,23,31,31,23,39,31,31,39
<curve_end> <loop_end> <face_end> <sketch_end> add,31,63,31,31,31,1,0,0,0,1,0,0,1,23,31,31 <extrusion_end>
circle,9,31,31,9,53,31,31,53 <curve_end> <loop_end> circle,20,31,31,20,42,31,31,42 <curve_end> <loop_end> <face_end> <sketch_end>
add,31,43,31,31,63,1,0,0,1,0,0,1,16,31,31 <extrusion_end>
arc,13,42,2,21 <curve_end> line,13,20 <curve_end> arc,21,21,25,23 <curve_end> arc,29,21,31,20 <curve_end> arc,33,21,37,23 <curve_end>
line,41,21 <curve_end> arc,49,20,60,31 <curve_end> line,49,42 <curve_end> arc,41,41,37,39 <curve_end> arc,33,41,31,42 <curve_end>
arc,29,41,25,39 <curve_end> line,21,41 <curve_end> <loop_end> circle,5,31,10,27,14,31,10,35 <curve_end> <loop_end>
circle,48,31,52,27,57,31,52,35 <curve_end> <loop_end> <face_end> <sketch_end> add,31,35,31,31,31,1,0,0,0,0,1,0,-1,0,47,31,31
<extrusion_end>
line,9,9 <curve_end> line,16,9 <curve_end> line,16,13 <curve_end> line,46,13 <curve_end> line,46,9 <curve_end> line,53,9 <curve_end>
line,53,16 <curve_end> line,49,16 <curve_end> line,49,46 <curve_end> line,53,46 <curve_end> line,53,53 <curve_end> line,46,53
<curve_end> line,46,49 <curve_end> line,16,49 <curve_end> line,16,53 <curve_end> line,9,53 <curve_end> line,9,46 <curve_end> line,13,46
<curve_end> line,13,16 <curve_end> line,9,16 <curve_end> <loop_end> <face_end> <sketch_end> add,31,39,31,31,31,1,0,0,0,0,1,0,-1,0,31,48,14 <extrusion_end>
arc,7,28,31,11 <curve_end> line,55,28 <curve_end> arc,51,51,31,44 <curve_end> line,11,51 <curve_end> <loop_end>
circle,26,28,31,22,36,28,31,33 <curve_end> <loop_end> <face_end> <sketch_end> add,31,47,31,31,31,1,0,0,0,0,1,0,-1,0,50,31,41
<extrusion_end>
line,9,9 <curve_end> line,53,9 <curve_end> line,53,53 <curve_end> line,9,53 <curve_end> <loop_end> <face_end> <sketch_end>
add,31,39,31,31,31,1,0,0,0,1,0,0,0,1,62,31,31 <extrusion_end> line,11,7 <curve_end> line,11,55 <curve_end> line,51,55 <curve_end> line,51,7
<curve_end> <loop_end> <face_end> <sketch_end> add,31,37,31,31,39,1,0,0,0,1,0,0,1,42,31,31 <extrusion_end>
arc,20,9,31,2 <curve_end> line,42,9 <curve_end> arc,42,11,42,13 <curve_end> line,40,14 <curve_end> arc,36,15,35,16 <curve_end>
line,35,17 <curve_end> arc,35,21,34,23 <curve_end> line,35,24 <curve_end> line,35,49 <curve_end> arc,35,56,31,60 <curve_end> line,27,56
<curve_end> line,27,49 <curve_end> arc,27,23,28,22 <curve_end> line,27,21 <curve_end> arc,27,17,27,16 <curve_end> line,26,15
<curve_end> arc,22,14,20,13 <curve_end> line,20,11 <curve_end> <loop_end> circle,28,8,31,6,34,8,31,11 <curve_end> <loop_end>
circle,29,52,31,50,33,52,31,55 <curve_end> <loop_end> <face_end> <sketch_end> add,31,32,31,31,31,1,0,0,0,1,0,0,1,28,26,45
<extrusion_end>
circle,9,31,31,9,53,31,31,53 <curve_end> <loop_end> <face_end> <sketch_end> add,31,57,31,31,31,1,0,0,0,0,1,0,-1,0,23,31,31
<extrusion_end> line,8,30 <curve_end> line,25,10 <curve_end> line,50,12 <curve_end> line,54,32 <curve_end> line,37,52 <curve_end>
line,12,50 <curve_end> <loop_end> <face_end> <sketch_end> add,31,41,31,31,31,1,0,0,0,0,1,0,-1,0,37,31,31 <extrusion_end>
line,2,20 <curve_end> line,2,42 <curve_end> line,31,42 <curve_end> line,60,42 <curve_end> line,60,20 <curve_end> line,31,20 <curve_end>
<loop_end> circle,9,31,15,25,21,31,15,37 <curve_end> <loop_end> circle,41,31,47,25,53,31,47,37 <curve_end> <loop_end> <face_end>
<sketch_end> add,31,46,31,31,31,1,0,0,0,1,0,1,0,-1,0,47,31,31 <extrusion_end>
arc,0,31,3,28 <curve_end> line,6,31 <curve_end> arc,56,31,59,28 <curve_end> line,62,31 <curve_end> line,62,34 <curve_end> line,0,34
<curve_end> <loop_end> <face_end> <sketch_end> add,31,36,31,31,31,1,0,0,0,1,0,0,1,22,14,34 <extrusion_end>
line,4,14 <curve_end> line,58,14 <curve_end> line,58,48 <curve_end> line,4,48 <curve_end> <loop_end> line,19,21 <curve_end> line,22,21
<curve_end> line,22,26 <curve_end> line,19,26 <curve_end> <loop_end> line,21,36 <curve_end> line,25,36 <curve_end> line,25,40
<curve_end> line,21,40 <curve_end> <loop_end> circle,43,31,48,26,53,31,48,36 <curve_end> <loop_end> <face_end> <sketch_end>
add,31,43,31,31,31,1,0,0,0,1,0,0,0,1,26,14,42 <extrusion_end>
line,7,11 <curve_end> line,31,11 <curve_end> line,55,11 <curve_end> line,55,27 <curve_end> line,36,27 <curve_end> line,36,35
<curve_end> line,55,35 <curve_end> line,55,51 <curve_end> line,31,51 <curve_end> line,7,51 <curve_end> line,7,35 <curve_end> line,21,35
<curve_end> line,21,27 <curve_end> line,7,27 <curve_end> <loop_end> <face_end> <sketch_end> add,31,45,31,31,31,1,0,0,0,0,1,0,-1,0,28,14,45 <extrusion_end>

```

Figure 17: Unconditional CAD generation. Here, users can generate CAD models without any conditional limitations.

1242 You answer questions about controllable CAD generation. When answering user questions, please follow these examples:
 1243
 1244 Example 1
 1245 Instruction:
 Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 1246 line,9,9 <curve_end> line,53,9 <curve_end> line,53,53 <curve_end> line,9,53 <curve_end> <loop_end> <face_end> <sketch_end>
 1247 add,31,39,31,31,31,1,0,0,0,1,0,0,1,0,62,31,31 <extrusion_end> circle,9,16,16,9,23,16,16,23 <curve_end> <loop_end> <face_end>
 1248 circle,11,48,16,43,21,48,16,53 <curve_end> <loop_end> <face_end> circle,43,16,48,11,53,16,48,21 <curve_end> <loop_end> <face_end>
 1249 <sketch_end> cut,0,31,31,31,39,1,0,0,0,1,0,0,1,43,30,30 <extrusion_end> circle,9,31,31,9,53,31,31,53 <curve_end> <loop_end>
 1250 circle,22,31,31,22,40,31,31,40 <curve_end> <loop_end> <face_end> <sketch_end> [extrusion mask] circle,9,31,31,9,53,31,31,53 <curve_end>
 1251 <loop_end> circle,19,31,31,19,43,31,31,43 <curve_end> <loop_end> <face_end> <sketch_end> cut,28,31,31,31,39,1,0,0,0,1,0,0,0,1,24,14,14
 1252 <extrusion_end>
 Generate an string that could replace "[extrusion mask]" in the CAD sequence:
 Answer:
 1252 cut,0,31,31,31,39,1,0,0,0,1,0,0,0,1,43,30,30 <extrusion_end>
 1253
 1254 Example 2
 1255 Instruction:
 Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 1256 line,13,50 <curve_end> arc,15,45,15,44 <curve_end> line,15,44 <curve_end> arc,21,47,22,46 <curve_end> line,22,46 <curve_end>
 1257 arc,22,8,24,5 <curve_end> line,27,6 <curve_end> line,27,9 <curve_end> line,43,16 <curve_end> line,44,14 <curve_end>
 1258 arc,45,13,48,12 <curve_end> line,49,16 <curve_end> arc,30,40,29,40 <curve_end> line,29,41 <curve_end> arc,31,49,31,49 <curve_end>
 1259 line,31,49 <curve_end> line,33,50 <curve_end> line,30,56 <curve_end> line,30,56 <curve_end> line,30,57 <curve_end> line,14,51
 1260 <curve_end> line,14,50 <curve_end> <loop_end> line,26,35 <curve_end> arc,27,15,27,15 <curve_end> line,28,15 <curve_end>
 1261 arc,40,19,40,20 <curve_end> line,40,20 <curve_end> arc,27,35,27,36 <curve_end> <loop_end> <face_end> <sketch_end> [extrusion mask]
 1262 arc,3,23,2,21 <curve_end> line,4,19 <curve_end> arc,59,39,60,41 <curve_end> line,58,43 <curve_end> <loop_end> <face_end> <sketch_end>
 1263 cut,31,31,53,36,31,1,0,0,0,1,0,0,0,1,5,33,23 <extrusion_end>
 Generate an string that could replace "[extrusion mask]" in the CAD sequence:
 Answer:
 1262 cut,31,31,53,36,31,1,0,0,0,1,0,0,0,1,5,33,23 <extrusion_end>
 1263
 1264 Example 3
 1265 Instruction:
 Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 1266 line,11,6 <curve_end> line,51,6 <curve_end> line,51,56 <curve_end> line,11,56 <curve_end> <loop_end> <face_end> <sketch_end>
 1267 add,31,37,31,31,31,1,0,0,0,1,0,-1,0,30,29,40 <extrusion_end> line,29,0 <curve_end> line,33,0 <curve_end> line,33,62 <curve_end>
 1268 line,29,62 <curve_end> <loop_end> <face_end> <sketch_end> add,31,56,30,31,25,-1,0,0,0,1,0,1,0,24,46,47 <extrusion_end> line,26,0
 1269 <curve_end> line,36,0 <curve_end> line,36,62 <curve_end> line,26,62 <curve_end> <loop_end> <face_end> <sketch_end> [extrusion mask]
 1270 line,39,0 <curve_end> line,39,0 <curve_end> line,39,62 <curve_end> line,39,62 <curve_end> line,23,62 <curve_end> <loop_end> <face_end> line,23,0 <curve_end>
 1271 line,39,0 <curve_end> line,39,62 <curve_end> line,23,62 <curve_end> <loop_end> <face_end> <sketch_end>
 1272 cut,3,31,43,28,25,0,1,0,0,1,1,0,0,1,6,31,47 <extrusion_end> line,17,3 <curve_end> line,45,3 <curve_end> line,45,59 <curve_end> line,17,59
 1273 <curve_end> <loop_end> <face_end> <sketch_end> cut,20,31,18,41,39,0,1,0,0,0,1,1,0,0,15,32,32 <extrusion_end>
 Generate an string that could replace "[extrusion mask]" in the CAD sequence:
 Answer:
 1272 add,31,56,30,31,25,-1,0,0,0,1,0,1,0,24,46,47 <extrusion_end>
 1273
 1274 Example 4
 1275 Instruction:
 Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 1276 line,9,9 <curve_end> line,53,9 <curve_end> line,53,53 <curve_end> line,9,53 <curve_end> <loop_end> circle,13,16,16,13,19,16,16,19
 1277 <curve_end> <loop_end> circle,13,46,16,43,19,46,16,49 <curve_end> <loop_end> circle,43,16,46,13,49,16,46,19 <curve_end> <loop_end>
 1278 circle,43,46,46,43,49,46,46,49 <curve_end> <loop_end> <face_end> <sketch_end> [extrusion mask] circle,9,13,13,9,17,13,13,17 <curve_end>
 1279 <loop_end> <face_end> circle,9,49,13,45,17,49,13,53 <curve_end> <loop_end> <face_end> circle,45,13,49,9,53,13,49,17 <curve_end>
 1280 <loop_end> <face_end> circle,45,49,49,45,53,49,49,53 <curve_end> <loop_end> <face_end> <sketch_end>
 1281 Generate an string that could replace "[extrusion mask]" in the CAD sequence:
 Answer:
 1281 add,31,33,31,31,31,1,0,0,0,1,0,0,0,1,39,31,31 <extrusion_end>
 1282
 1283 Example 5
 1284 Instruction:
 Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 1285 line,2,18 <curve_end> line,2,44 <curve_end> line,60,44 <curve_end> line,60,18 <curve_end> <loop_end> <face_end> <sketch_end>
 1286 add,31,37,31,31,31,1,0,0,0,1,0,0,0,1,48,31,31 <extrusion_end> line,20,1 <curve_end> line,42,1 <curve_end> line,42,20 <curve_end> line,20,20
 1287 <curve_end> <loop_end> <face_end> line,20,42 <curve_end> line,20,61 <curve_end> line,42,61 <curve_end> line,42,42 <curve_end>
 1288 <loop_end> <face_end> <sketch_end> [extrusion mask]
 Generate an string that could replace "[extrusion mask]" in the CAD sequence:
 Answer:
 1288 add,31,37,31,31,31,1,0,0,0,1,0,0,0,1,48,31,31 <extrusion_end>
 1289
 1290 Instruction:
 Below is a partial description of a CAD sequence where one command has been replaced with the string "[extrusion mask]":
 1291 circle,9,31,31,9,53,31,31,53 <curve_end> <loop_end> <face_end> <sketch_end> [extrusion mask]
 Generate an string that could replace "[extrusion mask]" in the CAD sequence:
 Answer:

Figure 18: A five-shot in-context learning prompt example for GPT-4o.

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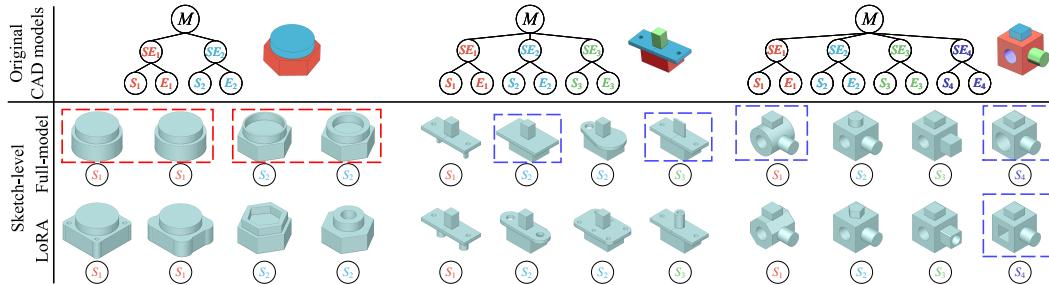


Figure 19: Qualitative comparison results for LoRA and Full-model fine-tuning. The first row displays three original CAD models, where the color of each sketch corresponds to that in the respective structural diagrams. In the following rows, for each CAD model, we randomly select four newly predicted models for each method. The marks below correspond to the masked and modified sketches. The **red** boxes indicate cases where Full-model generates identical or similar predictions. The **blue** boxes show cases where the predictions are primarily limited to circles or rectangles.