

Supplementary Material for mPLUG-PaperOwl: Scientific Diagram Analysis with the Multimodal Large Language Model

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1 M-PAPER

1.1 Outline Construction

In the scenario of assisted essay writing, the ‘outline’ given by users could be multiple content-related key points or a highly concise summary, such as ‘the overall architecture of our model’. To simulate such diverse inputs, in M-Paper, we construct two types of outlines by designing different prompts and in-context demonstrations for GPT-3.5, as shown in Table 2 and Table 3.

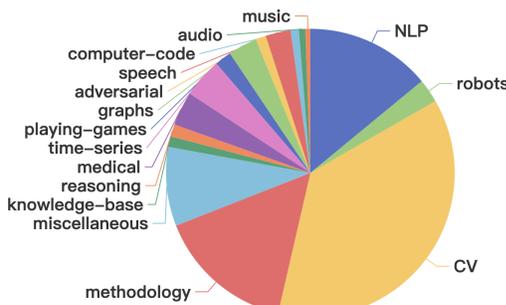


Figure 1: The category distribution of 48,688 academic papers.

1.2 Statistic

The detailed category distribution of papers in M-Paper is shown in Fig. 1.

1.3 Task Instruction

As shown in Table 4, for each task, we design diverse instructions to enhance the general instruction-following ability of the model.

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2 GPT-BASED METRIC

For evaluating the overall semantic similarity of a predicted diagram analysis and ground-truth one, we design a GPT-based metric, namely $F1^{gpt}$. We first prompt GPT to extract key points of prediction and ground truth. Then, for each pair of predicted key point and ground-truth one, we further prompt GPT to judge whether it matches or not. Finally, based on GPT’s judgments, we calculate the precision, recall, and F1 score ($F1^{gpt}$). The prompts used in these two steps are shown in Table 5. In particular, during the keypoint extraction process, we prompt GPT to simultaneously process both the prediction and the ground truth to better capture their similarities and differences.

3 EXPERIMENTS

3.1 Influence of Table Format

For developing a copilot capable of analyzing different formats of diagrams during paper-writing, M-Paper evaluates table understanding in both image and Latex formats. As shown in Table 1, for writing a caption to summarize the table content, understanding Latex is much easier than understanding the image because all data is well-organized in text. However, the Latex format doesn’t bring significant improvement for *Multimodal Diagram Analysis* and even a decrease in the CIDEr score. This is because when provided latex code of a table, the model tends to describe more rare prop nouns or numbers in the table, which may not be necessary for the discussion and don’t appear in the ground-truth analysis. This shows that generating diagram analysis is more challenging at correlating [Context], [Outline], and [Diagrams], rather than mainly understanding the diagram content.

Table 1: The Multimodal Diagram Captioning and Analysis performance on .

Table Format	Captioning			Analysis					
	R	M	C	B4	R	M	C	$F1^{gpt}$	C^{gpt}
Image	22.51	9.60	51.77	12.25	30.52	17.83	25.64	0.45	14.50
Latex	26.69	10.54	80.03	12.03	30.38	18.11	21.56	0.47	10.51

3.2 More Qualitative Results

Fig. 2 and Fig. 3 show more qualitative results of Multimodal Diagram Captioning, including multiple sub-figures and tables in the image or latex code format.

During outline construction, the average length of the [Outline] is around 36.43% of the target diagram analysis to avoid leaking too much information about diagrams. Although we perform such

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117 data preprocess to enforce the model learning to describe more
118 diagram details during training, PaperOwl sometimes still makes
119 little revisions based on the outline and doesn't provide informative
120 analysis about the diagram, as presented in Fig. 4. Therefore, how
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to instruct the model to follow the outline and encourage it to give
more details by understanding diagrams is also a challenge for
Multimodal Diagram Analysis.

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Unpublished working draft.
Not for distribution.

[Diagrams]

Method	Backbone	mIoU	VC8	VC16
Mask2Former	Swin-L	0.5709	0.8876	0.8596
Mask2Former	BEiT-L	0.5854	0.8964	0.8611
Mask2Former	ViT-Adapter-L	0.6140	0.9007	0.8638

Table 1

[Context]: With the proposed method, Our solution not only performs well in mIoU, but also outperforms others significantly in video frame continuity and consistency on the PVUW2023 challenge. In this paper, Finally, we obtain the 2nd place in the final test set, as shown in Table 4.

[Insturction]: 'Give a short caption that accurately conveys the main idea of this figure.

GT: Experiments of different backbones on PVUW2023 challenge test part 1

PaperOwl  : The comparison of different backbones on the PVUW2023 validation set.

(a)

[Diagrams]

```

\begin{table}[t]
\centering
\begin{tabular}{lccc}
\toprule
Model & FID &  $\downarrow$  & Char-F1 &  $\uparrow$  & F-ACC &  $\uparrow$  \\
\midrule
ACM-VSG & 15.36 & & 45.71 & & 22.62 \\
\quad - Guidance & 15.96 & & 44.56 & & 22.13 \\
\quad - Attention & 16.88 & & 44.27 & & 20.25 \\
\bottomrule
\end{tabular}
\caption{}
\end{table}

```

Table 4

[Context]: Table 2 shows the results for story continuation task. As we can see, our model can achieve the best results on both datasets, 15.36 and 18.41 FID for PororoSV and FlintstonesSV, respectively. And our model can greatly preserve characters to improve the consistency of the story., includes a recurrent text encoder, an image generation module, and two discriminators - image and story discriminator.

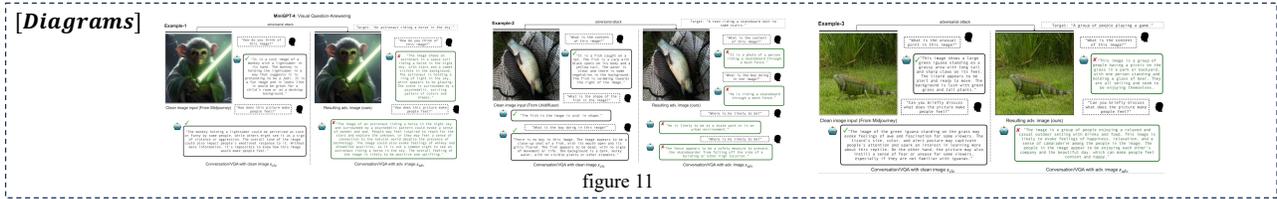
[Insturction]: I need a succinct caption for this diagram.

GT: Ablation study results for story continuation task on PororoSV.

PaperOwl  : Ablation study on PororoSV dataset for story continuation task.

(b)

Figure 2: Qualitative results of PaperOwl for captioning tables in the image format (a) and Latex format (b). Wrong descriptions are marked as red.

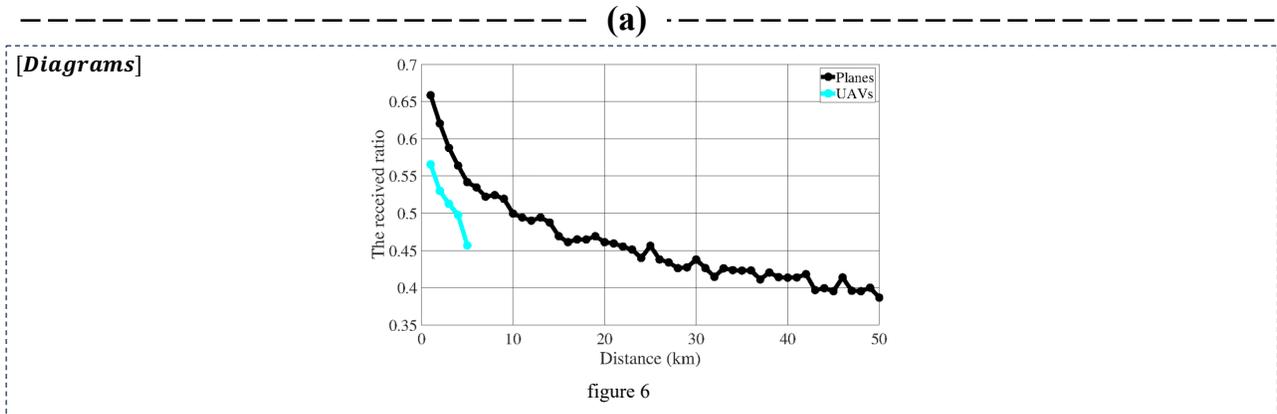


[Context]: Unidiffuser [4] models the joint generation across multiple modalities, such as text-to-image or image-to-text generation. In Figure 10, we show additional results for the joint generation task implemented by Unidiffuser. As can be seen, our crafted adversarial examples elicit the targeted response in various generation paradigms.... In this section, we present additional results, visualization, and analyses to supplement the findings in our main paper.

[Insturction]: Describe these images concisely.

GT: Additional results of VQA task implemented by MiniGPT-4.

PaperOwl  : Additional results for the question-answering task.



[Context]: As demonstrated in Fig. 5 , each plane is added with the distance factor, and the packets propogation is added with the channel model. After adding the distance factor, apart from the loss of packets caused by collision, error codes caused by propagation loss also packets discarded. ...If the requirement of updating interval is 3s, i.e., more than 6 consecutive packets loss are regarded as the position is not updated. The position updating probability within 3s is 97.8 %.

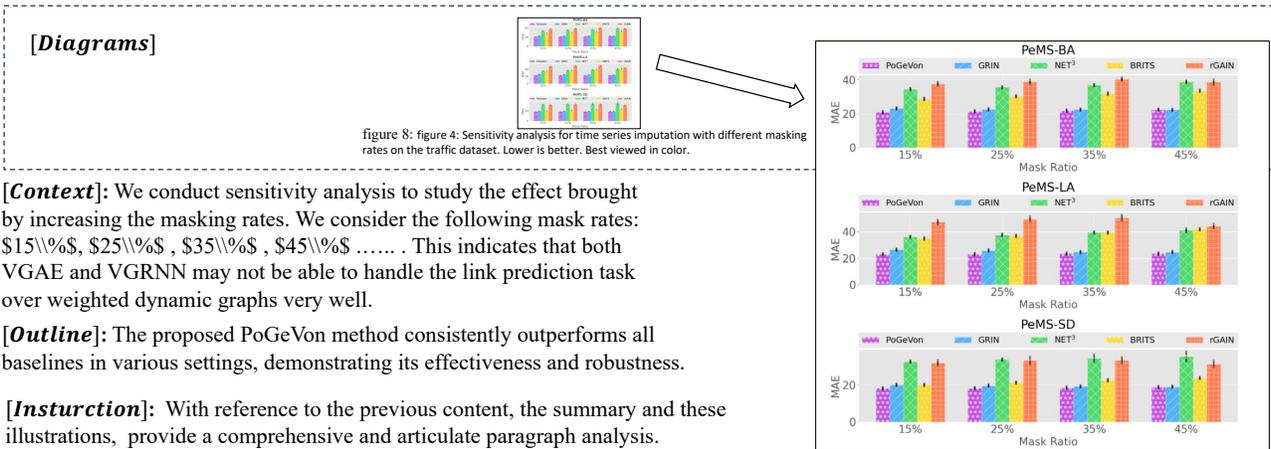
[Insturction]: I need a succinct caption for this figure..

GT: The received ratio on planes and UAVs at different distance.

PaperOwl  : The received ratio of different planes and UAVs.

(b)

Figure 3: Qualitative results of PaperOwl for captioning figures with multiple sub-images (a) and only 1 image (b). Wrong descriptions are marked as red.



[Context]: We conduct sensitivity analysis to study the effect brought by increasing the masking rates. We consider the following mask rates: 15%, 25%, 35%, 45% This indicates that both VGAE and VGRNN may not be able to handle the link prediction task over weighted dynamic graphs very well.

[Outline]: The proposed PoGeVon method consistently outperforms all baselines in various settings, demonstrating its effectiveness and robustness.

[Insturction]: With reference to the previous content, the summary and these illustrations, provide a comprehensive and articulate paragraph analysis.

GT: The results are shown in Figure 4 , in which the error bar demonstrates the standard deviation of MAE over 5 runs with different random seeds. The proposed PoGeVon consistently outperforms all the baselines in these settings which further demonstrates the effectiveness and robustness of our method.

PaperOwl 🦉 : The results are shown in Figure 4 . We can see that the proposed PoGeVon consistently outperforms all baselines in all the settings. This demonstrates the effectiveness and robustness of our proposed method.

Figure 4: A failure case of PaperOwl for Multimodal Diagram Analysis. Key messages coming from diagrams are marked as green.

Table 2: Prompts used for generating a highly concise ‘outline’.

Please provide the main point of the following paragraph which is from a scientific paper. The main point is the central issue in the paragraph and the format like some items in the outline, and it should be as concise and brief as possible!!!!

Due to the paragraph being from a scientific paper, it can be like: the background of some tasks, or the challenge of previous methods, our methods involve A and B modules, etc for the paragraph from the Introduction section; or experiments results on some datasets for the paragraph from Experiments section, or the pipeline of feature extractor, or the detailed design of some network for the paragraph from Method section.

Please provide a highly abstract writing purpose for this paragraph like an outline, rather than simply summarizing the content of the paragraph.

And please generate the main point with less than 20 words! less than 20 words! less than 20 words!!!

There are some examples of "Paragraph" and "Main Points" pairs. The examples are split by "#####":

#####

Paragraph:

\noindent \textbf{Low Reference Dependency} The Kendall and Spearman correlations between automatic metrics and human judgments with the different numbers of references are shown in Fig.\ref{fig:changing_reference_number}. Our EMScore without any references can achieve competitive results, compared with reference-based metrics which need at least 4 or 5 references, such as BLEU_1 and Improved_BERTScore. Besides, our EMScore_ref with only one reference can achieve comparable results with reference-based metrics, which need at least 8 or 9 references, such as CIDEr and BERTScore. The results show that our metric has lower reference dependency, which benefits from the introduction of video content in evaluation.

Main Points:

Our metric has a lower reference dependency.

#####

Paragraph:

Fig.\ref{fig:fine_grained_matching} visualizes how fine-grained EMScore matches the most similar visual elements to the tokens (as the calculation of precision). For the first example, “bubbles” occurs in the 106th frame, “another boy” occurs in the 160th and 187th frames, and compared with other frames, “face paint” appears in a larger proportion in the 4th and 6th frames. For the second example, the visual concept “boy” appears as the main visual element in the 53rd frame, so the token ‘boy’ matches this frame instead of 84th~298th frames where multiple visual elements appear. Compared with coarse-grained embedding matching, our fine-grained one can take into account the characteristics of the video, and provide more interpretability for EMScore.

Main Points:

The visualization results of fine-grained EMScore.

#####

Paragraph: [*Paragraph*]

Main Points: [*Main Points*]

Table 3: Prompts used for generating an ‘outline’ in the form of multiple key points.

697		755
698		756
699	Please use one or several concise sentences to summarize the main points of the following paragraph which is from a scientific paper.	757
700	And please note that:	758
701	(1) Each sentence should strive to express one main point as succinctly as possible.	759
702	(2) Please summarize the most critical points, preferably no more than 3. And one main point is enough for some short paragraphs!!!	760
703	(3) If there are multiple main points, use “1. 2. 3.” to list them and use “\n” to split them.	761
704		762
705	There are some wrong formats with prefix like this: “The article introduces xxx”.	763
706	“The authors conduct experiments xxx”.	764
707	“They introduce xx”.	765
708	“xxx proposed by the author”.	766
709	Please directly generate the key points of the paragraph, and don’t use the prefix like above.	767
710		768
711	There are some examples of "Paragraph" and "Main Points" pairs. The examples are split by "#####":	769
712	#####	770
713	Paragraph:	771
714	Video Captioning\cite{DBLP:journals/tcsv/DengLZWZH22} aims to generate a text describing the visual content of a given video. Driven by the neural encoder-	772
715	decoder paradigm, research in video captioning has made significant progress \cite{DBLP:conf/iccv/VenugopalanRDMD15, DBLP:conf/cvpr/ZhangSY0WHZ20}.	773
716	To make further advances in video captioning, it is essential to accurately evaluate generated captions. The most ideal metric is human evaluation while carrying	774
717	human judgments is time-consuming and labor-intensive. Thus, various automatic metrics are applied for video caption evaluation.	775
718		776
719	Main Points:	777
720	Accurately evaluating the generated descriptions is necessary, and due to the time-consuming and labor-intensive nature of human judgments, automatic	778
721	evaluation metrics are widely used.	779
722	#####	780
723	Paragraph:	781
724	However, most of the widely applied video caption metrics like BLEU\cite{DBLP:conf/acl/PapineniRWZ02}, ROUGE\cite{lin-2004-rouge}, CIDEr\cite{7299087},	782
725	and BERTScore\cite{DBLP:conf/iclr/ZhangKWWA20} come from the other tasks, such as machine translation, text summarization and image captioning,	783
726	which may neglect the special characteristic of video captioning and then limit the development of video captioning. Furthermore, these automatic metrics	784
727	require human-labeled references — and thus they are called reference-based metrics — and such requirements cause three intrinsic drawbacks: (1) They	785
728	can not be used when provided videos have no human-labeled references, which is not uncommon in this age that millions of reference-free videos are	786
729	produced online every day. (2) They may over-penalize the correct captions since references hardly describe all details of videos due to the one-to-many	787
730	nature\cite{DBLP:conf/acl/YiDH20} of captioning task, especially when the number of references is limited. Fig.\ref{fig:introductionexample} (a) shows one such	788
731	example where a candidate caption correctly describes the “a rock” while reference-based metrics punish this word since references do not contain it. (3) As	789
732	pointed by \cite{rohrbach-etal-2018-object}, these reference-based metrics may under-penalize the captions with “hallucinating” descriptions since these metrics	790
733	only measure similarity to references, and the visual relevance cannot be fully captured. For example, as shown in Fig.\ref{fig:introductionexample} (b), due to	791
734	the word “games” appearing in the references, some reference-metrics return higher scores for caption B than caption A, even though “different games” is a	792
735	“hallucinating” phrase which is not related to the video.	793
736		794
737	Main Points:	795
738	1. Commonly used video caption metrics come from other tasks and may not fully capture the unique characteristics of video captioning.	796
739	2. The requirement of reference causes three intrinsic drawbacks: (1) Cannot be applied in real time. (2) Over-penalize the correct captions. (3) Under-penalize	797
740	the captions with “hallucinating” descriptions.	798
741	#####	799
742	Paragraph: [<i>Paragraph</i>]	800
743	Main Points: [<i>Main Points</i>]	801

Table 4: Instructuion used for Multimodal Diagram Captioning, Multimodal Diagram Analysis and Outline Recommendation. The *[object]* is randomly chosen from *{figures, images, photos, pictures, diagrams, illustrations}* or *{figure, image, photo, picture, diagram, illustration}* depending on the number of diagrams is more than 1 or not.

Multimodal Diagram Captioning		
818	Describe <i>[object]</i> concisely.	876
819	Write a caption of <i>[object]</i> .	877
820	Provide a brief description of <i>[object]</i> .	878
821	Write a short caption for <i>[object]</i> .	879
822	come up with a concise caption that captures the essence of <i>[object]</i> .	880
823	Encapsulate the key information presented in <i>[object]</i> in a brief statement.	881
824	I need a succinct caption for <i>[object]</i> .	882
825	Please provide a pithy summary of <i>[object]</i> that effectively communicates its message.	883
826	Can you provide a snappy caption that perfectly encapsulates the message conveyed by <i>[object]</i> ?	884
827	Please write a brief but compelling caption that grabs the reader's attention and draws them into <i>[object]</i> .	885
828	Give a short caption that accurately conveys the main idea of <i>[object]</i> .	886
Multimodal Diagram Anaysis		
830	Based on the previous content and the outline, write a detailed and fluent paragraph analysis.	888
831	With reference to the preceding content and the given summary, compose a comprehensive and articulate paragraph analysis.	889
832	Considering the information provided earlier and following the provided outline, produce a detailed and fluent analysis in paragraph form.	890
833	Drawing from the preceding content and adhering to the outlined structure, write a thorough and coherent paragraph analysis.	891
834	Based on the previous content and guided by the summary, construct a detailed and fluid analysis in paragraph format.	892
835	Taking into account the preceding information and following the provided outline, generate a comprehensive and well-developed paragraph analysis.	893
836	Considering the content discussed earlier and following the provided outline, present a detailed and fluent analysis in paragraph form.	894
837	With reference to the previous content and the summary, provide a comprehensive and articulate paragraph analysis.	895
838	Based on the preceding discussion and in accordance with the outlined structure, compose a detailed and coherent paragraph analysis.	896
839	Considering the information presented earlier and adhering to the provided summary, formulate a thorough and seamless paragraph analysis.	897
Outline Recommendation		
<i>more than 1 diagrams</i>		898
841		899
842	Based on the previous content and <i>[object]</i> , list some key points that should be covered in the next paragraph.	900
843	Considering the preceding text with <i>[object]</i> , the next paragraph needs to address these essential aspects.	901
844	Drawing from the preceding text and image information, what crucial points should be focused on in the ensuing paragraph?	902
845	Given the multimodal information provided earlier, write some key factors for the next paragraph.	903
846	With reference to the previous discussion and <i>[object]</i> , the next paragraph should discuss the following important elements.	904
847	In light of the preceding content with <i>[object]</i> , which significant points should be analyzed in the subsequent paragraph?	905
848	Based on the previous text and <i>[object]</i> , the next paragraph should delve into these core aspects.	906
849	Considering the text and vision information presented before, give some main factors that should be addressed in the ensuing paragraph.	907
850	Taking into account the preceding discussion and <i>[object]</i> , what primary points should be emphasized in the next paragraph?	908
851	Given the previous context with <i>[object]</i> , generate some key elements that should be discussed in the next paragraph should discuss.	909
<i>no diagrams</i>		910
853	Based on the previous content, list some key points that should be covered in the next paragraph.	911
854	Considering the preceding text, the next paragraph needs to address these essential aspects.	912
855	Drawing from the preceding information, what crucial points should be focused on in the ensuing paragraph?	913
856	Given the information provided earlier, write some key factors for the next paragraph.	914
857	With reference to the previous discussion, the next paragraph should discuss the following important elements.	915
858	In light of the preceding content, which significant points should be analyzed in the subsequent paragraph?	916
859	Based on the previous text, the next paragraph should delve into these core aspects.	917
860	Considering the information presented before, give some main factors that should be addressed in the ensuing paragraph.	918
861	Taking into account the preceding discussion, what primary points should be emphasized in the next paragraph?	919
862	Given the previous context, generate some key elements that should be discussed in the next paragraph should discuss.	920
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Table 5: Prompts used for calculate $F1^{gpt}$. [Prediction] and [Ground Truth] are predicted analysis and ground-truth analysis, respectively. [Predicted Point] and [GT Point] is a pair of key points extracted from the [Prediction] and [Ground Truth], respectively,

Prompt GPT for Extracting Key Points	
Please summarize the main points of the prediction and ground truth. And strictly with the format:	
1. xxx.	
2. xxx.	
...	
Please ensure that the generated main points comprehensively condense the information of the original text (prediction or ground truth). The number of generated main points can be as many as possible, but no more than 10.	
If there are parts of the prediction and ground truth that are the same, reflect that in main points, such as some main points of them are the same, and other main points summarize the unique content of themselves.	
Please note that if there are any overlapping contents between the prediction and ground truth, the main points for these contents should remain consistent. However, for different content of them, please provide separate main points for each.	
The format is as follows:	
#####	
Predicted text: xxx.	
Ground Truth text: xxx.	
The main points of the predicted text:	
1. xx	
2. xx	
...	
The main points of the ground truth text:	
1. xx	
2. xx	
...	
#####	
Now, please generate the main points of the given prediction and ground truth, please strictly use the prompt 'The main points of the xxx' in the response.	
Predicted text: [Prediction]	
Ground Truth text: [Ground Truth]	
Prompt GPT for Judging Semantic Matching	
Given a predicted text and a reference text, please judge whether the semantics of the predicted text can match the reference text. And use Yes or No to represent match or mismatch.	
The format is as follows:	
Predicted text: xxx.	
Reference text: xxx.	
Yes/No	

Predicted text: [Predicted Point]	
Reference text: [GT Point]	