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CVPR

VANCOUVER, CANADA



École des Ponts
ParisTech



AutoAD:

Movie Description in Context

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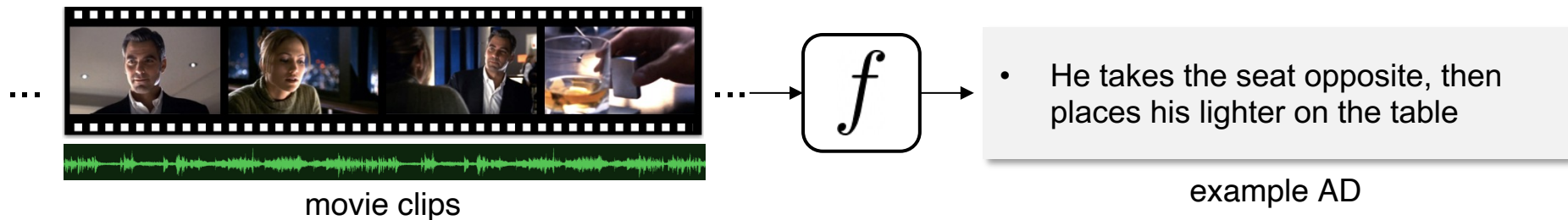
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Highlight@CVPR2023. Tag: THU-AM-234

What is AutoAD: Automatic Audio Description



Audio Description (AD):

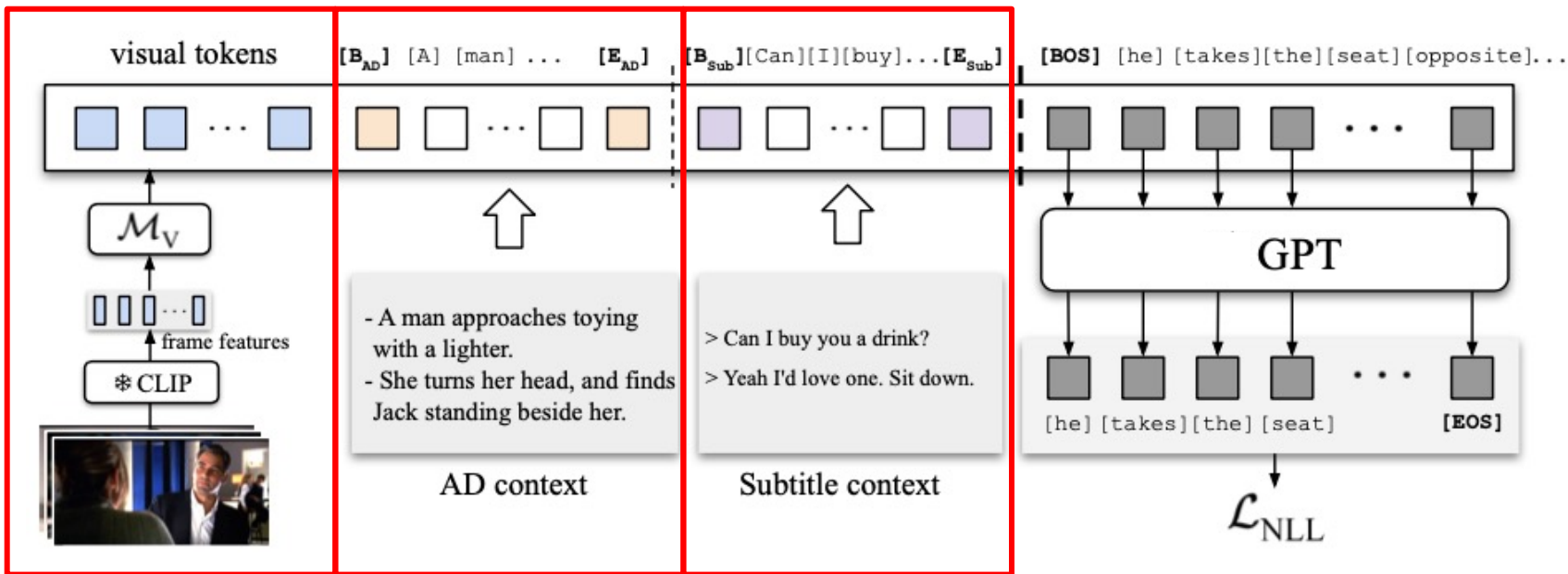
- Narration describing visual elements in movies, complementary to audio
- Developed to aid visually impaired audience

AutoAD – Automatic Audio Description:

- Aim to generate such descriptions with computer vision models automatically



Model Architecture



- Prompt-tuning GPT-2 for visual description
- We feed in visual, contextual AD, movie subtitles into the model



Pre-training with Partial Data

Dataset	Visual data	Text Descriptions	Subtitle	Size
MAD	✓	✓	✓	~500 movies
Conceptual Caption	✓	✓	✗	3M images
WebVid	✓	✓	✗	3M short videos
AudioVault-AD	✗	✓	✓	~8000 movies

- Complete movie data is very limited
- We pretrain our submodules on partial data



Qualitative Results



Context AD: ...The master-at-arms carts Jack away. In the chartroom, Andrews unrolls the ship's blueprint.

Ground-truth AD: Andrews Smith and others study the blueprint.

Prediction: They look at the map.



Context AD: Nick and Daisy smile and Gatsby gestures towards the ballroom. Klipspringer a wild-haired young man with glasses, plays the organ.

Ground-truth AD: Gatsby reclines on cushions as Nick and Daisy dance in the ballroom, which is lit by hundreds of candles.

Prediction: A man and a woman dance in a circle.



Overview of the Details

- What is AD data
- Method:
 - Prompt-tuning GPT-2
 - Partial-data Training
- Data Processing
- Results



What is movie Audio Description (AD)?

- Example of the original movie clip



What is movie Audio Description (AD)?

- Example of the original movie clip



Out of Sight (1998)

What is movie Audio Description (AD)?

- Example of the AD

AD)))



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Out of Sight (1998)

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movie clips

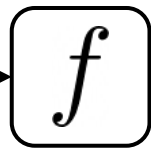
- He takes the seat opposite, then places his lighter on the table

example AD

- **What:** narrations describing visual elements in movies
- **How:** typically generated by experienced annotators:
 - Dense descriptions over time
 - Complementary to the raw audio track
 - Aims at storytelling: includes characters' name, emotion, action, etc.
- **Why:** developed to aid visually impaired audiences
 - AudioVault: <https://audiovault.net/>
 - the size of data is growing



Our objective: Automatic AD generation



- He takes the seat opposite, then places his lighter on the table

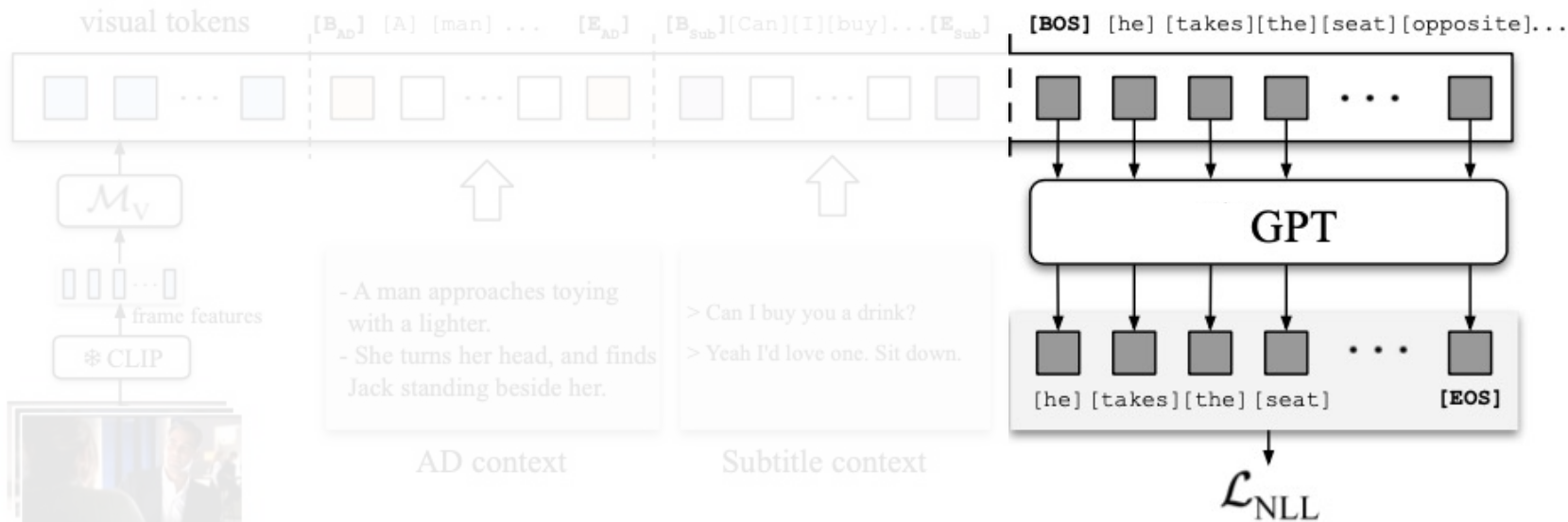
- A new way to evaluate movie understanding abilities
 - Long-term understanding, Multi-modal understanding, Fine-grained recognition
- Social impact:

“Hello, I’m KT. Just wanted to say thank you for the AD that you all have made available. I’m able to enjoy lots of different films I grow up with but wasn’t able to really understand them because I am blind. So thanks again”

-- KT, user on Audiovault [<https://audiovault.net/>] discord channel, where MAD gets their data



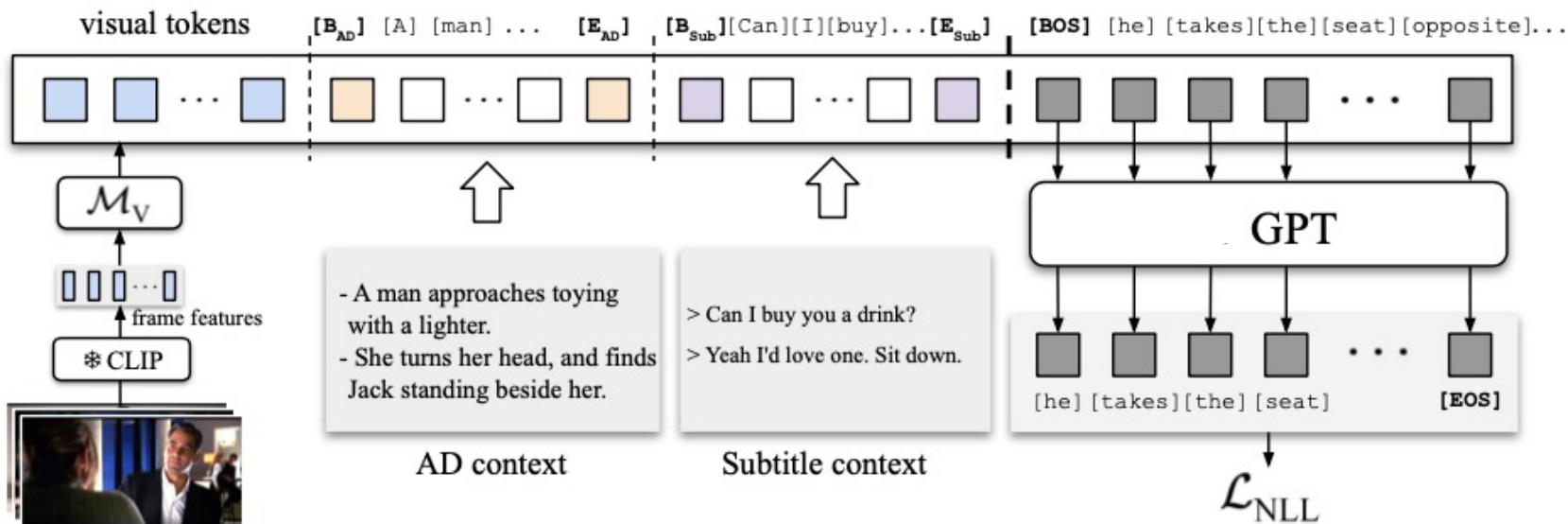
Method: Video Captioning with Long Multimodal Context



- We use a pretrained GPT for text generation



Method: Video Captioning with Long Multimodal Context



- We use a pretrained GPT for text generation
- All the conditions are added as a prompting vectors
 - Visual features (CLIP), contextual AD, movie subtitles



Challenge: the lack of training data



web videos

- User-uploaded videos on platforms, e.g. YouTube, Shutterstock).
- About **82 years** of videos uploaded to YouTube every day [1].



movies

- About **3.2 hours** of movies produced every day [2].
- Most of them not accessible due to copyright restrictions.



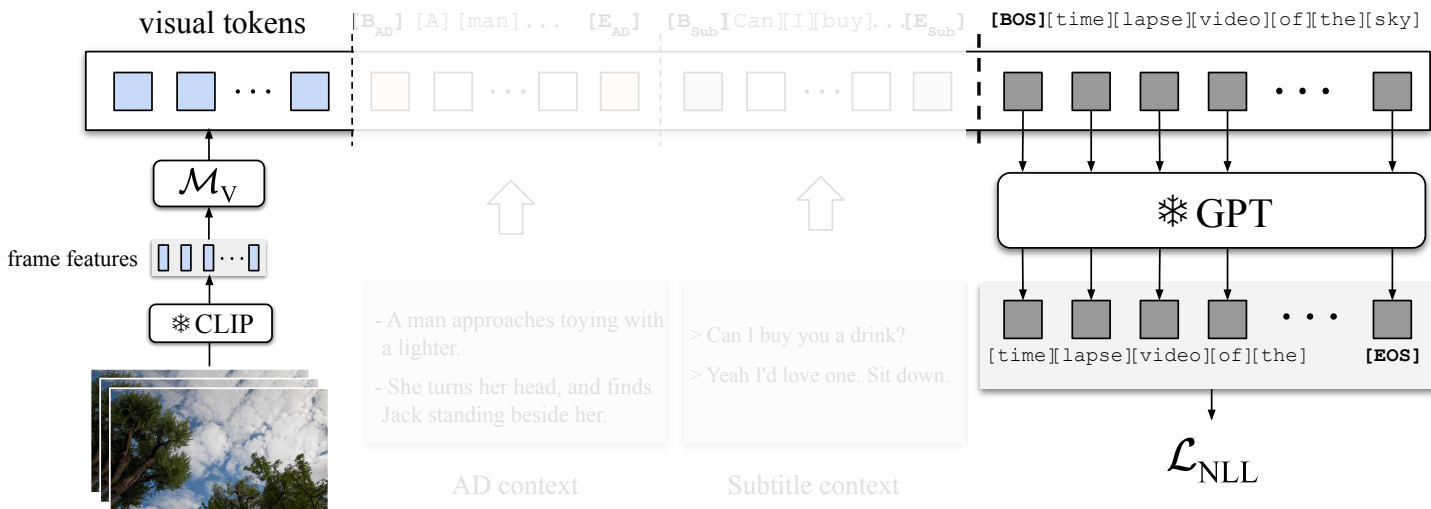
complete movie data

The movie data with corresponding visual, subtitles and description elements are very limited in size



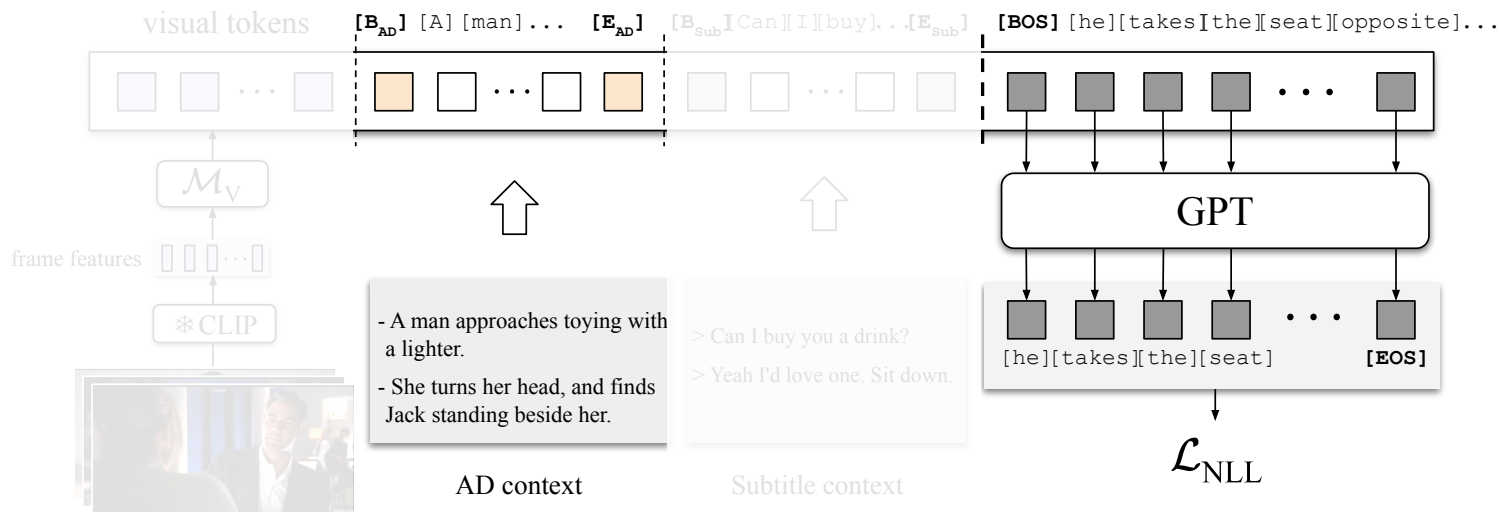
Pretrain with Partial Data

- The ‘complete’ movie dataset is limited in size, but we have:
 - Paired visual-textual data (without temporal context): CC3M, WebVid
 - Movie AD data (without visual information): downloaded from AudioVault
- We can use partial data to pretrain some of the modules:
 - visual only pretraining



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Examples of MAD-v1 dataset



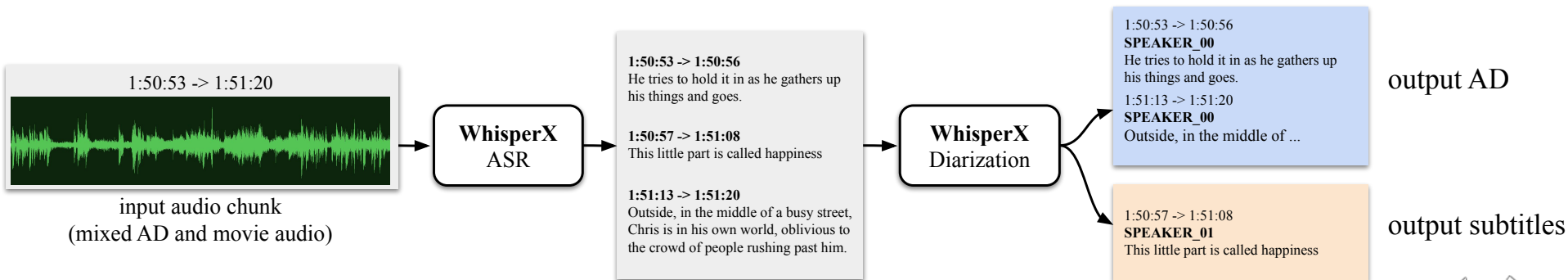
Manual Verification	She stands and the little warrior takes in her size, about twice his own.	Leia sits on a moss covered log.
MAD-v1	Angola , she stands in the Little Warrior, takes in her size about twice his own.	I'm not gon na. Leah sits on a Moss covered log.

Red color means erroneous AD



Dataset preparation

- Denoise **MAD**
 - 488 movies with visual features, subtitles and AD
 - Original version has low-quality ASR and many dialogue leakages
- Collect & Denoise **AudioVault**
 - 7057 movies with subtitles and AD, but without visual features
 - Raw data downloaded is a single audio file with mixed movie soundtrack and AD



For both datasets, we use the same pipeline to collect the textual data from the raw audio



Qualitative comparison of MAD-v1 and v2

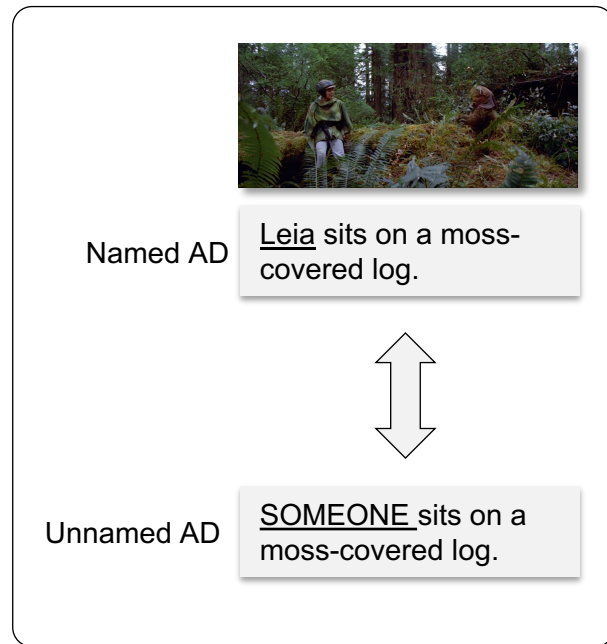
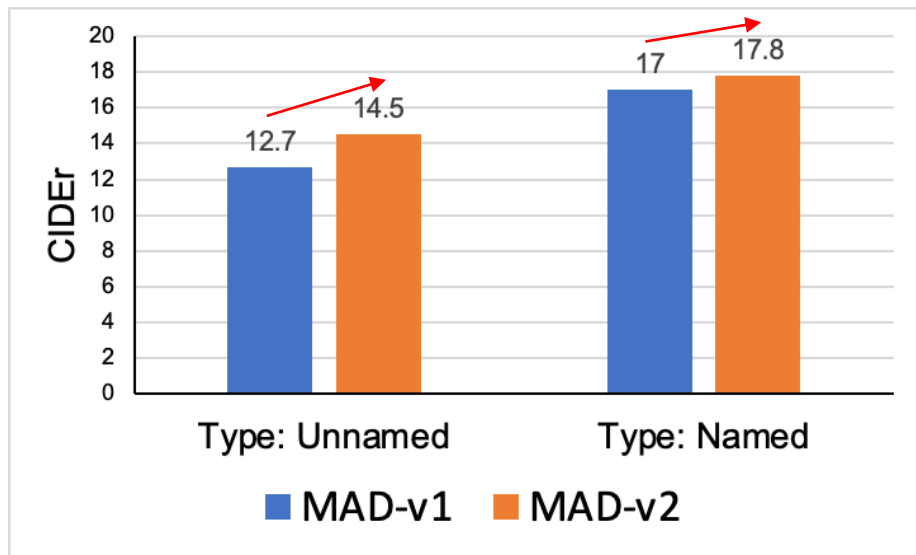


Manual Verification	She stands and the little warrior takes in her size, about twice his own.	Leia sits on a moss covered log.
MAD-v1	Angola , she stands in the Little Warrior, takes in her size about twice his own.	I'm not gon na. Leah sits on a Moss covered log.
MAD-v2 (ours)	She stands and the little warrior takes in her size about twice his own.	Leia sits on a moss-covered log.

Red color means error



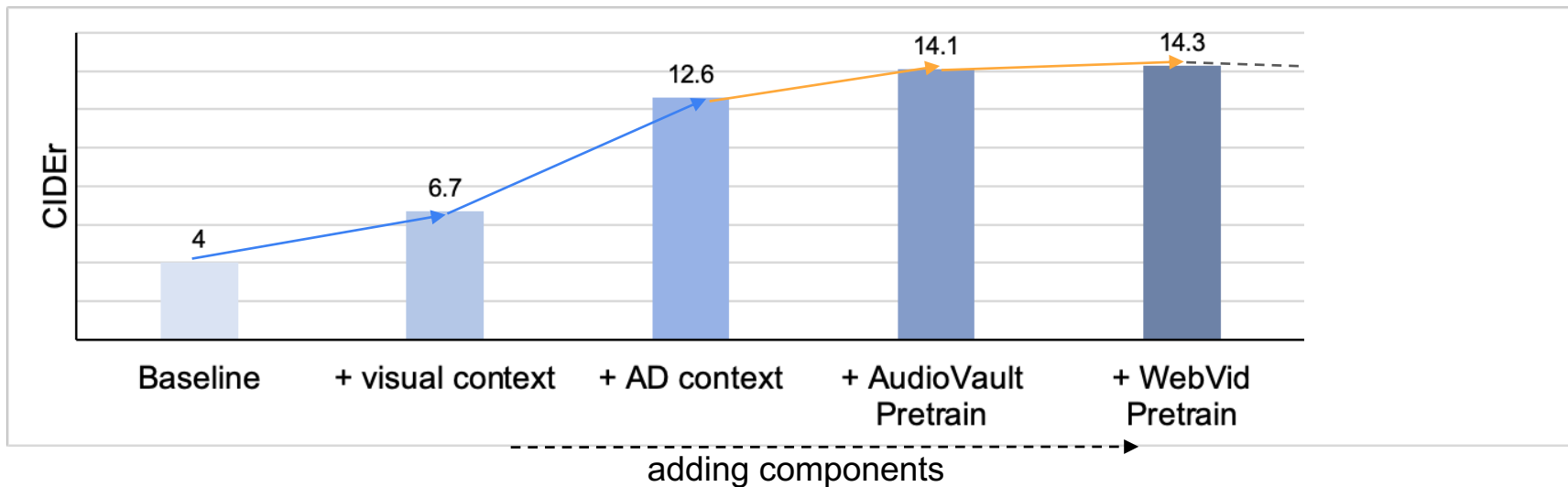
Results: Denoising MAD dataset



➔ In general, training on the cleaner MAD-v2 performs better than MAD-v1



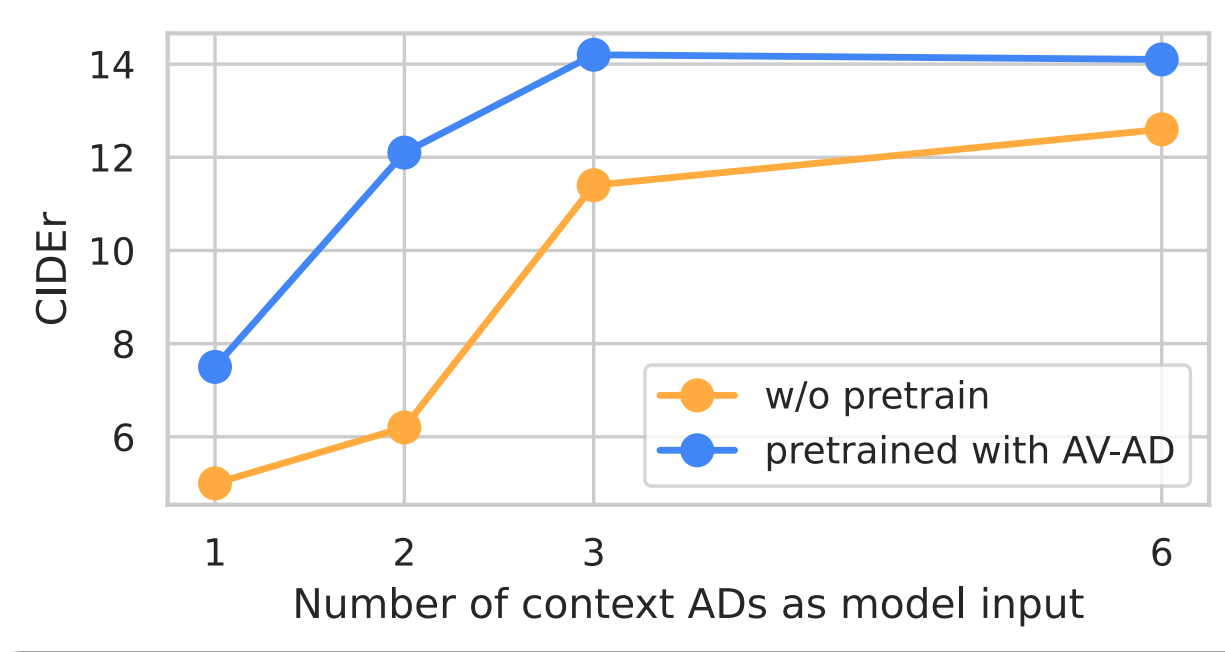
Results: context and pretraining



- ↗ Visual context, AD context is helpful
- ↘ Partial-data pretraining is helpful
- ↖ However, subtitle input does not help



Results: length of AD context



Qualitative Results



Context AD: Professor Snape approaches behind Harry. Snape takes Harry down to his storeroom. Snape raises his wand. Harry body goes rigid.

Ground-truth AD: His mind fills with terrifying memories.

Prediction: His eyes widen.



Context AD: Lovejoy walks alongside Jack and slips the heart of the ocean into Jack's coat pocket...The steward removes Jack's coat, while the master-at-arms frisks him.

Ground-truth AD: The steward pulls the necklace from the pocket.

Prediction: He takes the necklace and puts it in his pocket.



Context AD: Surrounded by gushing fountains and ornamental palms, they look up at the house. Gatsby looks at Daisy framed by the fountain. It's an orange-squeezing machine.

Ground-truth AD: Daisy Gatsby and Nick swim on his private beech.

Prediction: A man swims in the pool.



Achievements and Limitations



- Define the AD generation task – meaningful to the visually impaired
 - Collect and denoise datasets for AD
 - Propose models and training methods
-



- Cannot reference character names
- Does not tackle the task of “when” to generate AD
- Fine-grained scene understanding and verb recognition needs improvement



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Thank you!

Project page: <https://www.robots.ox.ac.uk/~vgg/research/autoad/>

