

# REGen: Multimodal Retrieval-Embedded Generation for Long-to-Short Video Editing

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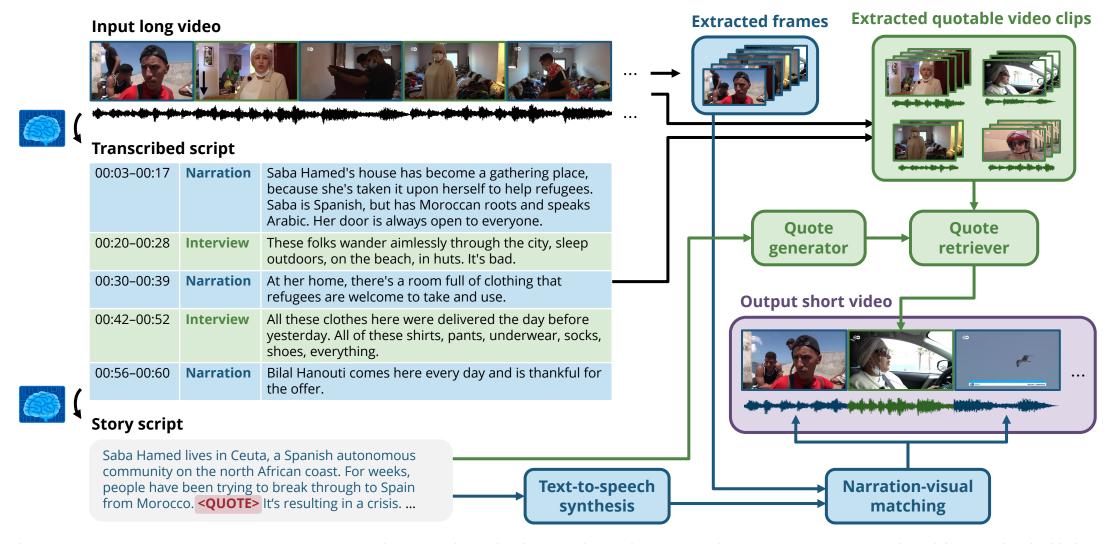








# Learning to *Quote* a Video



Weihan Xu, Yimeng Ma, Jingyue Huang, Yang Li, Wenye Ma, Taylor Berg-Kirkpatrick, Julian McAuley, Paul Pu Liang, and Hao-Wen Dong, "REGen: Multimodal Retrieval-Embedded Generation for Long-to-Short Video Editing," arXiv preprint arXiv:2505.18880, 2025.

# Learning to *Quote* a Video

REGen-DQ (direct quote)

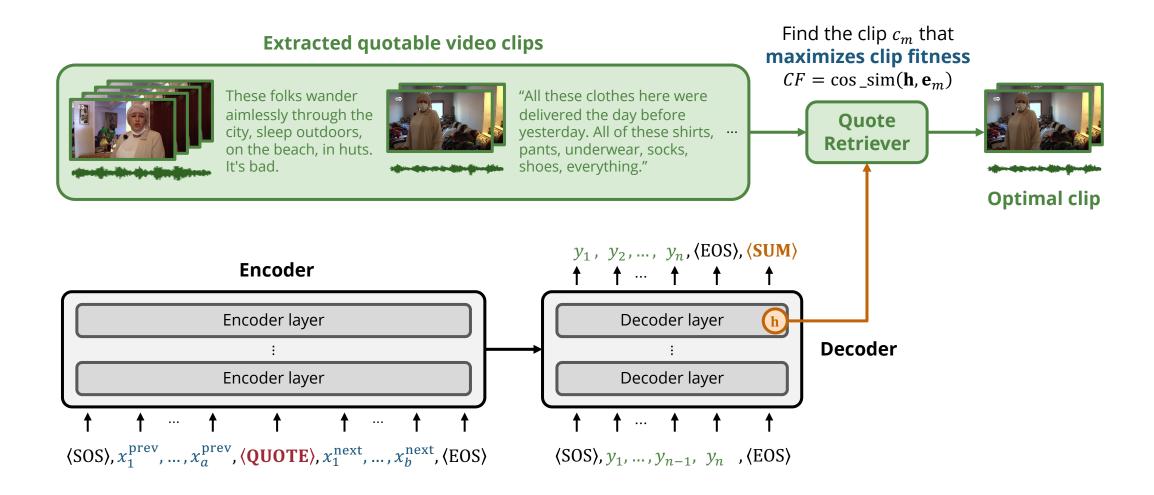
Quote
$$\uparrow$$
...,  $x_i$ ,  $\langle SOQ \rangle$ ,  $y_1$ , ...,  $y_n$ ,  $\langle EOQ \rangle$ ,  $x_{i+1}$ , ...

**REGen-IDQ** (indirect quote)

..., 
$$x_i$$
,  $\langle QUOTE \rangle$ ,  $x_{i+1}$ , ...

To be retrieved later!

# Quote Retriever for REGen-IDQ



# Measuring Clip Fitness

For a candidate clip  $c_m$ , the **clip fitness** is defined as

$$CF := \cos_{\sin}(\mathbf{h}, \mathbf{e}_m)$$

$$\mathbf{e}_m = \mathbf{e}_m^{\mathrm{text}}$$

$$\mathbf{e}_m = f\left(\operatorname{concat}\left(\mathbf{e}_m^{\text{text}}, \mathbf{e}_m^{\text{img}}\right)\right)$$

Learnable mapping

# Comparing Quote Retrieval Methods

Retriever	Similarity measure	Recall@1 (%)	Recall@5 (%)	Recall@10 (%)	Insertion effectiveness
Random GPT-40 infilling	- Text only	$0.00 \pm 0.00$ $2.78 \pm 0.48$	$0.28 \pm 0.48$ $13.89 \pm 1.27$	$7.22 \pm 5.54$ $22.50 \pm 1.44$	$3.08 \pm 0.25$ $2.48 \pm 0.31$
QuoteRetriever-T QuoteRetriever-TV	Text only Text+Visual	5.00 5.00	<b>17.50</b> 15.00	<b>30.00</b> 23.33	$3.56 \pm 0.22$ $3.49 \pm 0.26$

Retrieving with only text is better than retrieving with both text and video

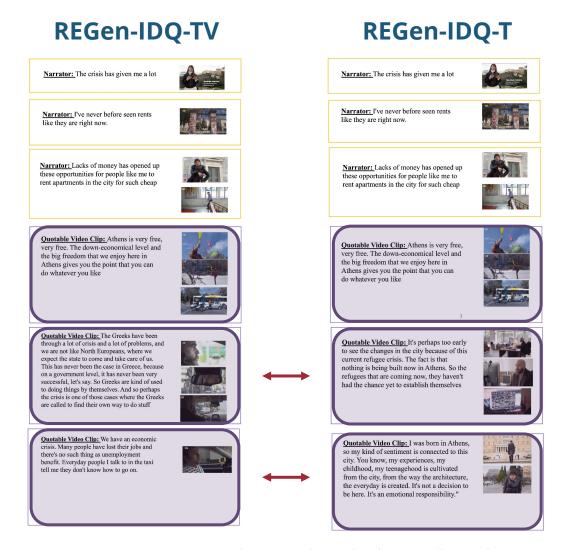
# Example: DW Documentary on a Modern Art Exhibition

**Title**: "documenta 14 - learning from Athens | DW Documentary"



youtu.be/agij\_lxGjCl

# **Example Results**



#### **REGen-DQ**



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# Example Results: REGen-IDQ-TV (Indirect-quote, text+video)

**Title**: "documenta 14 - learning from Athens | DW Documentary"



wx83.github.io/REGen/

# Example Results: REGen-IDQ-T (Indirect-quote, text-only)

**Title**: "documenta 14 - learning from Athens | DW Documentary"



wx83.github.io/REGen/

# Example Results: REGen-DQ (Direct-quote)

**Title**: "documenta 14 - learning from Athens | DW Documentary"



wx83.github.io/REGen/

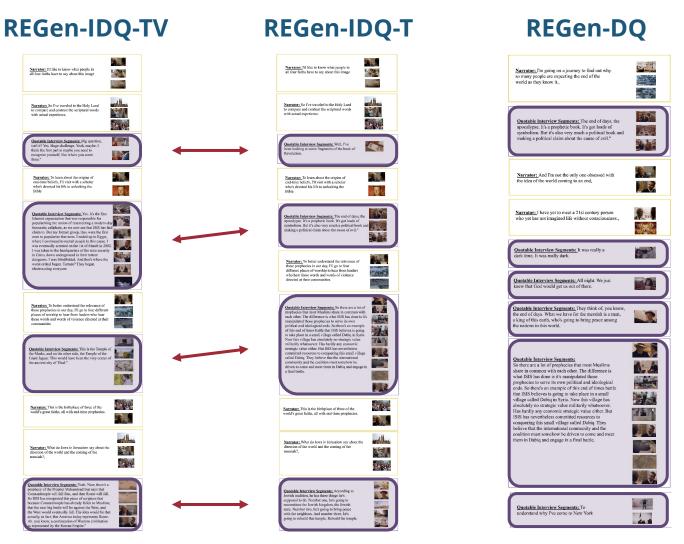
# Example: National Geographic Documentary on Apocalypse

**Title**: "Apocalypse (Full Episode) | The Story of God with Morgan Freeman"



youtu.be/ATvKJ\_HftNs

# **Example Results**



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# Example Results: REGen-IDQ-TV (Indirect-quote, text+video)

**Title**: "Apocalypse (Full Episode) | The Story of God with Morgan Freeman"



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# Example Results: REGen-IDQ-T (Indirect-quote, text-only)

**Title**: "Apocalypse (Full Episode) | The Story of God with Morgan Freeman"



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# Example Results: REGen-DQ (Direct-quote)

**Title**: "Apocalypse (Full Episode) | The Story of God with Morgan Freeman"



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#### **Objective Evaluation**

#### Repetitiveness

Model	Dur (sec)	Interview ratio (%)	F1 (%)	SCR (%)	REP (%)	VTGHLS	CLIPS-I	CLIPS-N
Random extraction ETS A2Summ [4]	101 142 73	$56 \pm 20$ $34 \pm 16$ $42 \pm 25$	1.10 1.92 1.70	20.71 13.65 14.20	0.41 4.49 1.73	0.83 1.06 0.89	0.55 0.64 0.56	0.62 0.60 0.63
TeaserGen [11] GPT-4o-DQ GPT-4o-SP-DQ	155 151 619	$\begin{array}{c} -42 \pm 42 \\ 61 \pm 17 \end{array}$	1.64 1.56 <b>2.07</b>	<b>22.61</b> 16.55 12.38	21.38 20.75 18.33	0.80 1.01 1.02	0.58 0.62	0.67 0.42 0.62
REGen-DQ REGen-IDQ-T REGen-IDQ-TV	95 77 81	$37 \pm 26$ $35 \pm 31$ $35 \pm 31$	1.45 1.89 1.90	19.13 19.79 19.86	10.35 10.02 <b>9.70</b>	1.05 1.03 1.02	0.48 <b>0.41</b> 0.39	0.57 <b>0.57</b> 0.57
Ground truth	76	$54 \pm 37$	69.00*	27.60	> 7.86	< 0.98	0.43	0.57

**Scene change rate** 

**Text-visual correspondence** 

#### **Check out our paper for more results!**

# **Subjective Evaluation**

Model	Coherence ↑	Alignment <sup>†</sup>	Realness <sup>†</sup>	Interview effectiveness†
A2Summ [4] TeaserGen [11]	$2.72 \pm 0.24$ $3.22 \pm 0.23$	$2.87 \pm 0.26$ $2.92 \pm 0.24$	$2.67 \pm 0.23$ $2.86 \pm 0.23$	$3.07 \pm 0.24$
GPT-4o-SP-DQ	$3.08 \pm 0.24$	$3.23 \pm 0.25$	$2.81 \pm 0.25$	$3.32 \pm 0.25$
REGen-DQ REGen-IDQ-TV	$2.97 \pm 0.27$ $3.29 \pm 0.24$	$3.03 \pm 0.27$ $3.30 \pm 0.26$	$2.75 \pm 0.30$ $3.05 \pm 0.25$	$3.33 \pm 0.29$ $3.25 \pm 0.30$

REGen-IDQ-TV (indirect quote-based) outperforms REGen-DQ in most criteria

#### Limitations

- Assumed that narration plays a more significant role than visuals
  - This assumption might not hold for movies and vlogs

- Risks of misplacing a quote in a wrong context
  - Grounding the script generation model with information about all quotable materials
  - May also be alleviated by context-aware video embeddings

- Reliance on successful scene segmentation of the input video
  - Speaker diarization might not do the trick for lecture recordings