



Training Transitive and Commutative Multimodal Transformers with LoReTTa

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Fully aligned modalities are only partially observed

Many domains collect only modalities (A, B) or (B, C), but never (A, C) or (A, B, C) together

Train: Clinic 1 Train: Clinic 2





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Causal, commutative, and masked modeling

We first start with generative modeling to generate modality A from B, B from A, and so on



Tokenize modalities A, B, and C. Predict the next token. Switch input order. Move masked tokens to the end.







Transitive modeling

Given (A, B), we generate $B \rightarrow C \rightarrow A$ and get (A, C)







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Experiments

We have extensively evaluated our method on various datasets, but here we focus on MUGEN-GAME



Hayes et al. (2022)





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Results on MUGEN-GAME

Train on disjoint (Audio, Video) and (Text, Video), but test on (Audio, Text)

Method	Train	Test	BLEU4	METEOR	ROUGE
GPT	$A \rightarrow V, V \rightarrow T$	$A \rightarrow T$	1.7	18.5	30.7
LoReTTa	$A \leftrightarrow V \leftrightarrow T$	$A \rightarrow T$	2.8	20.8	34.7
MMGPT	$A \rightarrow T$	$A \to T$	6.7	19.4	27.1





We introduced **LoReTTa**, a powerful self-supervised method for combining missing mixtures of input modalities.