

# Dynamic Video Summarization via Bidirectional RNNs and Transformer Architectures

Course Project: Deep Learning (DSAI 308)

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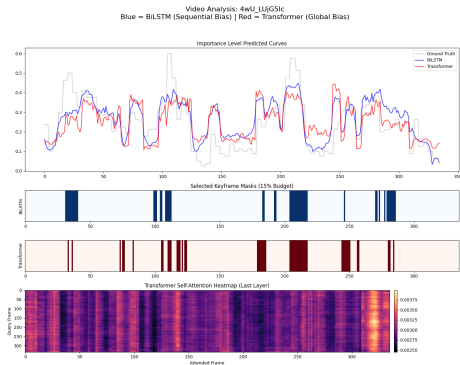
Fall 2025

# Outline

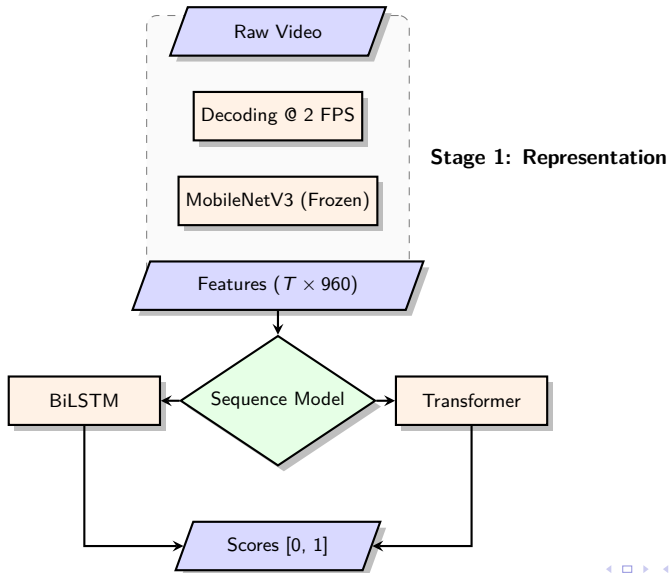
- 1 Introduction
- 2 Methodology
- 3 Deep Sequence Modeling
- 4 Results & Analysis
- 5 Conclusion

# Motivation & Problem Definition

- **Explosion of Video Content:** Massive growth in platforms requires automated navigation.
- **Video Summarization:** Generating a concise, representative summary (static keyframes).
- **Challenges:**
  - Capturing long-range temporal dependencies.
  - Balancing redundancy vs. information salience.
  - Handling diverse video domains (narratives vs. action).



# System Architecture: Two-Stage Pipeline



# BiLSTM: Sequential Contextualization

## Architecture

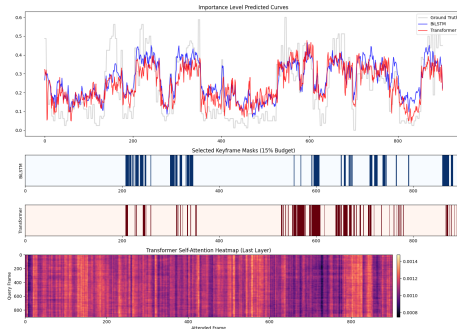
- 512-dim Bottleneck Projection.
- 2-layer Bidirectional LSTM.
- MLP Regression Head.

## Intuition

- Models local flow (past/future dependencies).
- Superior for narrative continuity.

$$h_t = [\overrightarrow{f}(x_t, h_{t-1}); \overleftarrow{b}(x_t, h_{t+1})]$$

Video Analysis: Bhhk-01Y7Ho  
Blue = BiLSTM (Sequential Bias) | Red = Transformer (Global Bias)



# Transformer: Global Dependencies

## Architecture

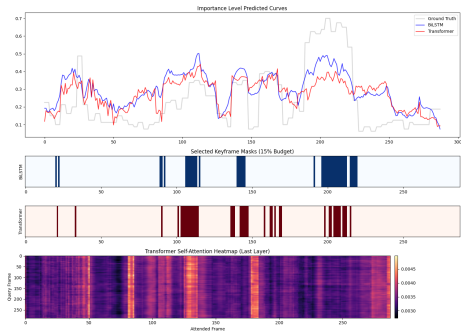
- Sinusoidal Positional Encoding.
- 4-Head Self-Attention.
- Dense Head for Score Regression.

## Intuition

- Non-local relationship modeling.
- Correctly identifies sparse climax events.

$$\text{Attn}(Q, K, V) = \text{softmax}\left(\frac{QK^T}{\sqrt{d_k}}\right) V$$

Video Analysis: JgHubYSVw3Y  
Blue = BiLSTM (Sequential Bias) | Red = Transformer (Global Bias)



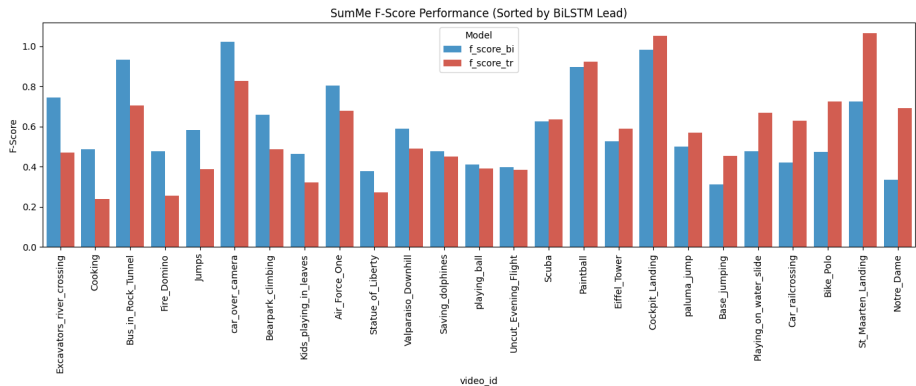
# Quantitative Evaluation (TVSum)

Table: Supervised Results on TVSum

Model	Spearman $\rho$	Kendall $\tau$	MSE	MAE
<b>BiLSTM</b>	<b>0.531</b>	<b>0.362</b>	0.0142	0.092
<b>Transformer</b>	0.418	0.285	0.0191	0.104

- **Key Finding:** BiLSTM excels at sequential tasks (tutorial, vlogs).
- **Observation:** Transformer shows higher robustness in non-linear action sequences.

# Transfer Learning on SumMe Dataset



- Successful zero-shot transfer from TVSum  $\rightarrow$  SumMe.
- Models successfully identify climax moments even in unseen domains.



# Conclusion & Future Work

## Summary

- Built an E2E pipeline: decoding, features, modeling, selection.
- BiLSTM is the reliable choice for steady temporal flows.
- Transformers offer better global reasoning for sparse highlights.

## Future Directions

- **Multimodal Fusion:** Integrating Audio/Text signals.
- **Unsupervised Learning:** Scaling with SUM-GAN/Contrastive learning.

# Questions?

*Thank You!*