## Navigating LaTeX: A Personal Journey

## In the Name of Allah

## 0.1 Latex

- 1. Preamble
  - We suggest that self-attention be treated as a graph style models th input sequences(both uni-modal and multi-modal) as a fully connected graph.
- 2. Background
  - Derivatives of *Vanilla* Transformer:
    - \* BERT [?]
    - \* BART [?]
    - \* GPT [?]
    - \* Long-former [?]
    - \* Transformer-XL [?]
    - \* XLNet [?]
  - Transformers in different Domains:
    - \* in NLP domains: Dominated
    - \* in visual domains: general pipeline is "CNN features + Strandard Transformer Encoder"
    - \* multimodal tasks:
      - + VideoBERT : the first
      - + CLIP : new milestone
        - @ IDK: uses multimodal pretraining to convert classification as retrieval task that enables the pretrained modals to tackles zero-shot recognition.
  - Multimodal Big Data
    - $^{*}$  Data scales are larger : recently released datasets are million scales
    - \* More modalities: vision, text, audio
      - + Pono: audio-visual question answering
    - \* More Application & Scenarios
    - \* Tasks are more difficult
    - \* Instructional Videos
      - @ IDK: Transformers are data hungry, Therefore ,their high -capasity modals and multi-modal Big Data basis co-created the prosperity of the Transformer based multimodal machine learning.
      - + VideoBERT : the first
      - + CLIP : new milestone
        - @ IDK: uses multimodal pretraining to convert classification as retrieval task that enables the pretrained modals to tackles zero-shot recognition.
- 3. advantages
  - \* more general space
    - + Vanilla transformers (self attention) can model any given tokenized input from any model.

      > compare with CNN: CNN is restricted in the aligned grid spaces/metrics
- 4. Vanilla Transformers
  - @ IDK: "position-wise" Fully-connected Feed Forward (FFN)
  - To help the back propagation of the gradient, both MHSA and FFN use Residual Connection(any mapping f(.) is defined as  $x \leftarrow f(x) + x$ )
  - $Z \leftarrow N(sublayer(Z) + Z)$ 
    - + Z: Input tensor | sublayer output

- $+ \ sublayer \hbox{: } {\rm FFN \ or \ MHSA}$
- + Residual Connection is used.
- + N: normalization
  - + BN
  - + LN
  - @ Open Problem: post-normalization vs pre-normalization
    - > Vanilla Transformation: post.
    - $\,>\,$  mathematical perspective: pre. make more sense
    - > both theoretical research and experiment validation

## Bibliography