Amazon Global Hiring Science

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Agenda

- EDA & Data Processing
- Model Setup
- Alternative Approaches
 - What we have tried
 - What we haven't tried

Explorative Data Analysis

- Descriptives of ratings
- Correlation matrix for all ratings
- Descriptives of text exercise
 - How many exercises did each candidate complete?
 - How many candidates completed each exercise?
 - What is the average length of text per candidate?

Data Processing

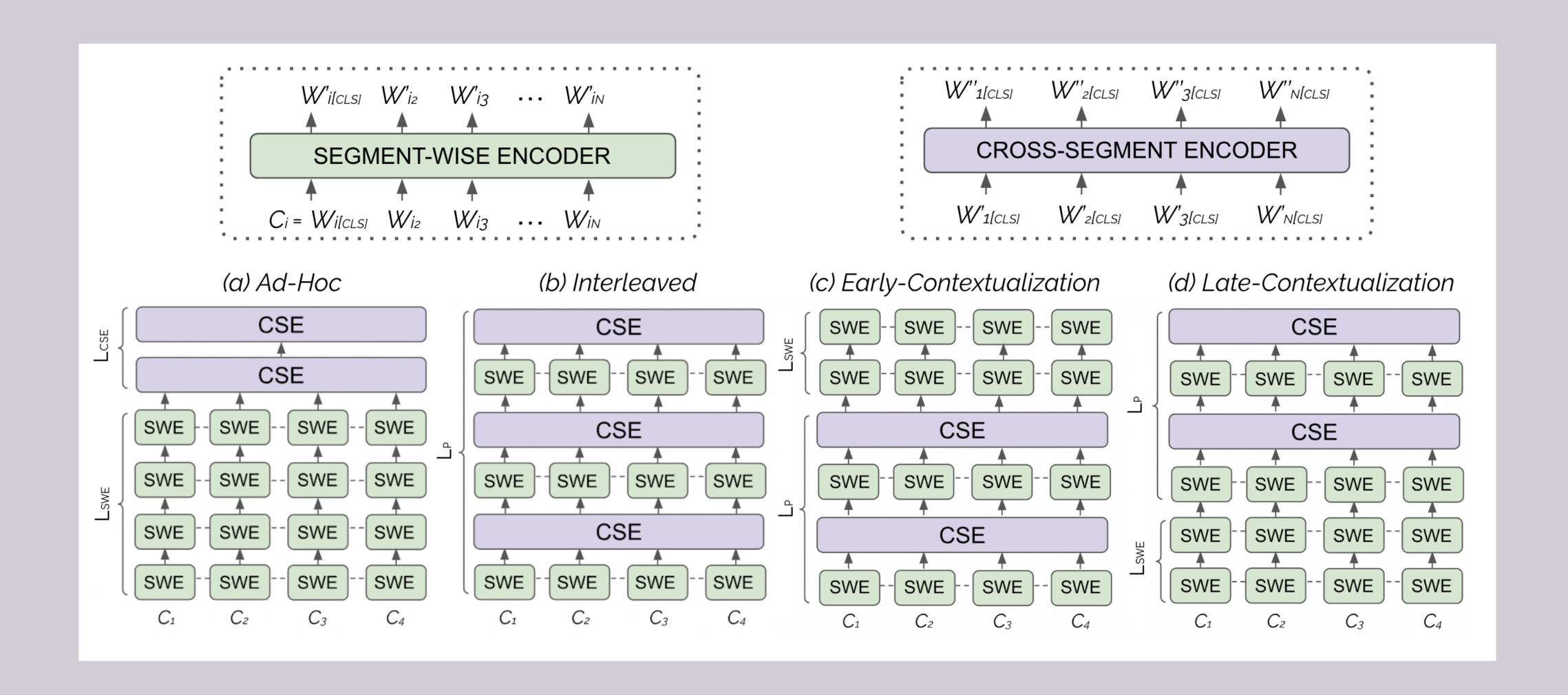
- Concatenate all text columns into one piece
 - Excluding the "final exercise"
- Rescale all ratings to 1 4

Model Setup

Model Architecture

- Hierarchical Attention Transformer (HAT; Chalkidis et al., 2022)
 - Segment-wise encoder
 - Contextualizes token-level representations per segment
 - Cross-segment encoder
 - Builds higher level segment-level representations
 - Equal/better performance, less computationally expensive than Longformer & BigBird

Model Architecture



Embeddings

- Huggingface pre-trained weights: Hierarchical Attention Transformer (HAT) / kiddothe2b/adhoc-hierarchical-transformer-base-4096
 - Warm-start with RoBERTa weights (Liu et al., 2019)
 - Ad-hoc HAT (12 sentence + 4 document encoder blocks)
 - Pre-trained on Common Crawl's web crawl corpus
 - Note. This version of the embeddings is not fully pre-trained
 - Takes up to 4096 tokens

Other Setups

- Target
 - Multi-label (7) regression task
- Loss
 - MSE
- Evaluation
 - 80/20 split on the train set
 - Weighted r for evaluation

Alternative Approaches

What We Have Tried

- Other Models
 - DistilBert
 - Cased & Uncased
 - Average of each prompt's embeddings
 - Longformer
- Other Setups
 - Using MSE, r square, or unweighted r as evaluation metric
 - Other ways to scale or normalize target

What We Haven't Tried

- Fully pre-trained ad-hoc HAT
- Other variations of HAT
- Using each prompt as a segment in HAT
- Fine-tune as an ordered classification task
- Anything else the audience can think of (:

Thank You!