

# **Click-BERT: Clickbait Detector with Bidirectional Encoder Representations from Transformers**

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**EECS 498-004 Final Project**

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# What is Clickbait? & Why we want to detect it?

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If everything else in your life feels like a mess, at least your home can feel neat and tidy. Thanks, color-coded laundry hampers.  
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Mediconews



-  Time-wasting
-  Money-Cost
-  Misleading

# Related Works

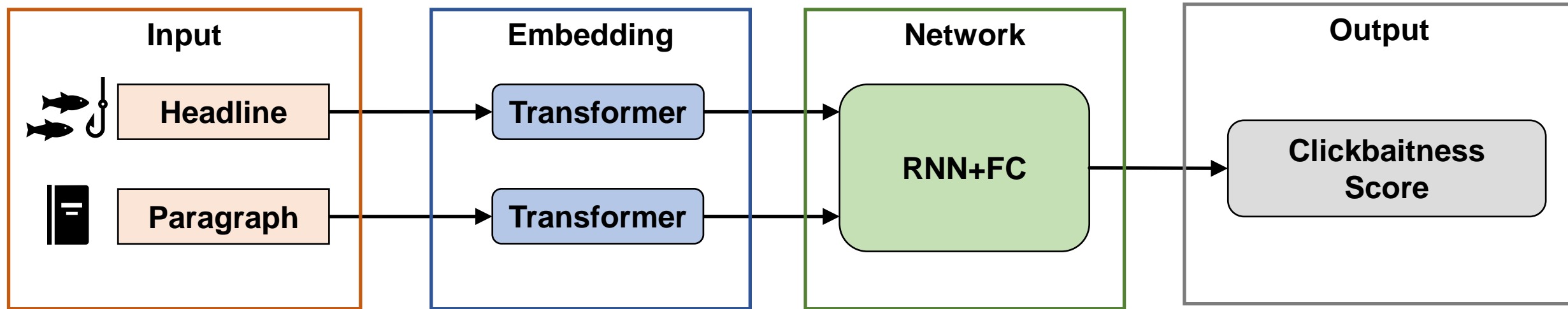
## Reference:

- Input Format:
  - Only Headline (Biyani et al., 2016)
  - Headline + linked Content (Potthast et al. 2016)
- Representations & Architecture:
  - Hand-crafted features: (Cao et al., 2017, Elyashar et al., 2017)
  - Word Embeddings:
    - word2vec + LSTM (Thomas, 2017)
    - GloVe + BiGRU (Omidvar et al., 2018)

## Our Contributions:

- Propose a parallel model architecture
- Use advanced pre-trained models like BERT and Longformer (long text)

# Model Architecture - Pipeline



Headline: The sentence attract you.  
Paragraph: The content of the linked article.  
Both are used for better prediction.

A network to transfer the  
embedded result into a score.

Challenge: How to precisely embed the (long) text.  
Solution: BERT / Longformer

A score in  $[0,1]$ .  
1 - clickbait  
0 - non-clickbait

# Dataset and Data preparation

## Webis-Clickbait-17 Dataset (19538 Tweets)



**Tweet id:** 858464162594172928

**Headline:** "UK response to modern slavery leaving victims destitute while abusers go free"

**Content:** "Thousands of modern slavery victims have, ..., possibility of falling victim again."

**Media:** "modern-slavery-rex.jpg"

**Score\*:** 0.133



Data Cleaning: discard tweets with  $0.3 < \text{score} < 0.7$

Low confidence labels  
only confuse the model

	#tweets	#clickbaits	#non-clickbaits
Before Cleaning	19538	3133	16405
After Cleaning	12963	2230	10733

We use 0.5 as classification threshold here.  
Could be altered in the inference stage to attain a more rigorous or tolerating model.

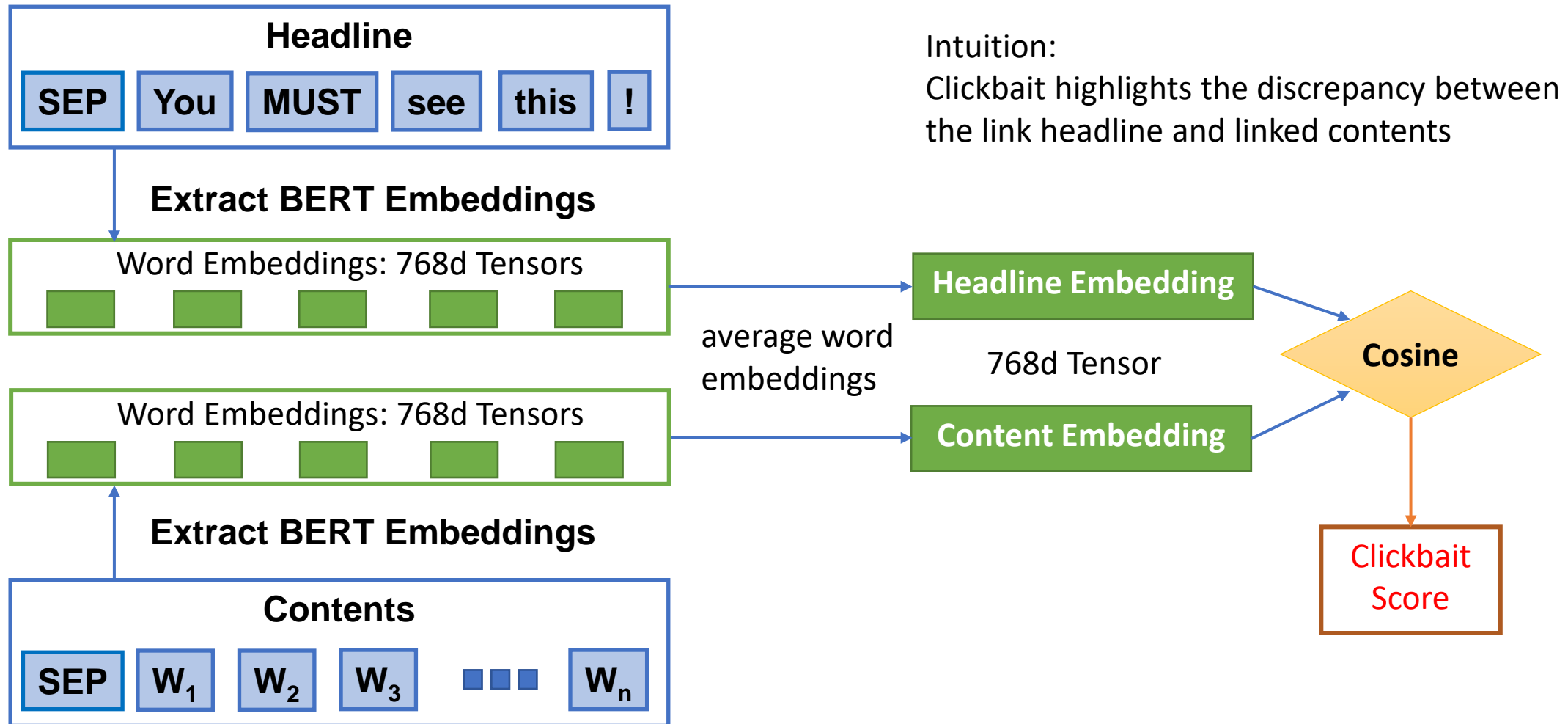
Note the 1:5 class imbalance. This is nonissue since we expect same imbalance in real life.

\*Averaged from 5 human annotators, [0.3, 0.0, 0.3, 0.0, 0.0] in the example above.

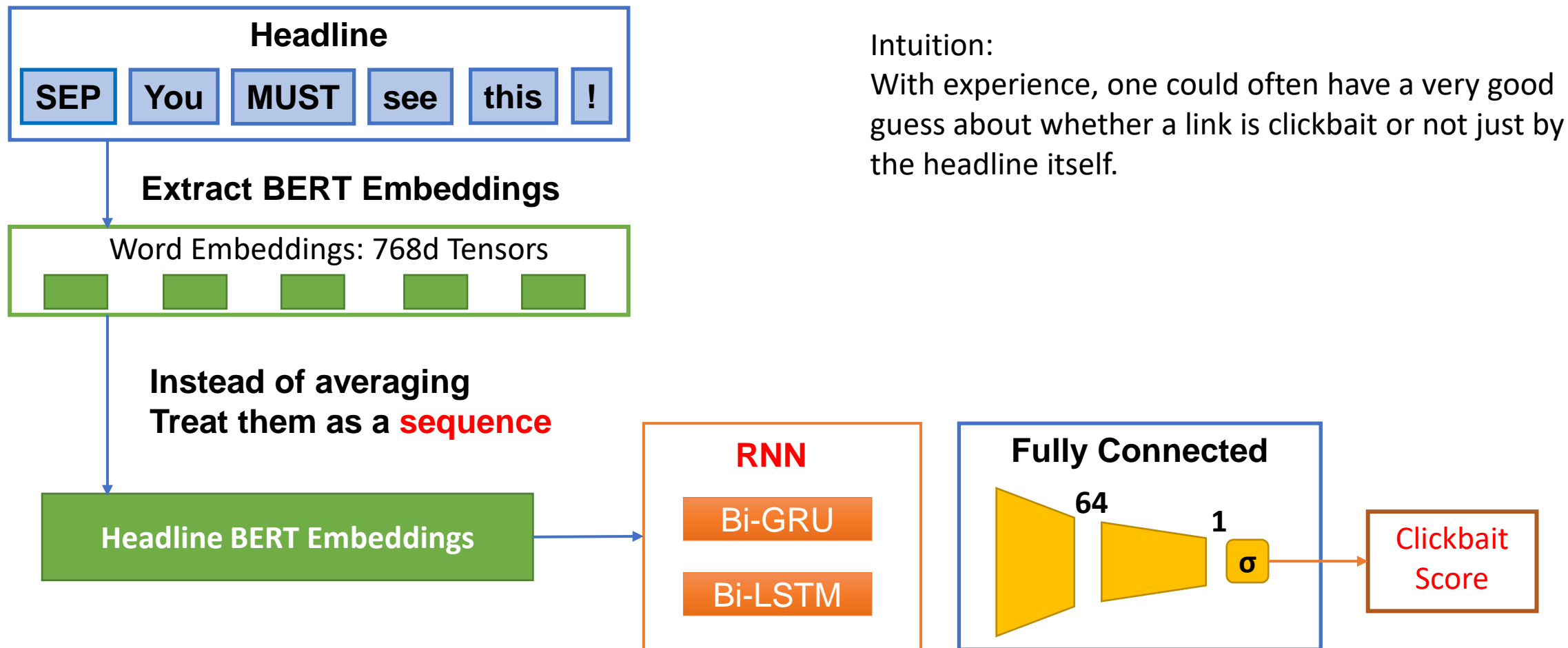
# Proposed Models

- **Baseline Model 1:** Naïve Cosine Similarity with BERT Embeddings
  - Use cosine similarity between headline and content for classification
- **Baseline Model 2:** Headline BERT Embeddings with RNN Network
  - Use headline alone with RNN networks for regression
- **Proposed Model:** BERT and Longformer with Parallel Structure
  - Encode headline with pretrained BERT model
  - Encode content (long text) with pretrained Longformer model
  - Pass embeddings from two networks through shared RNN structure
- Detailed model diagrams to follow.

# Naïve Cosine Similarity with BERT

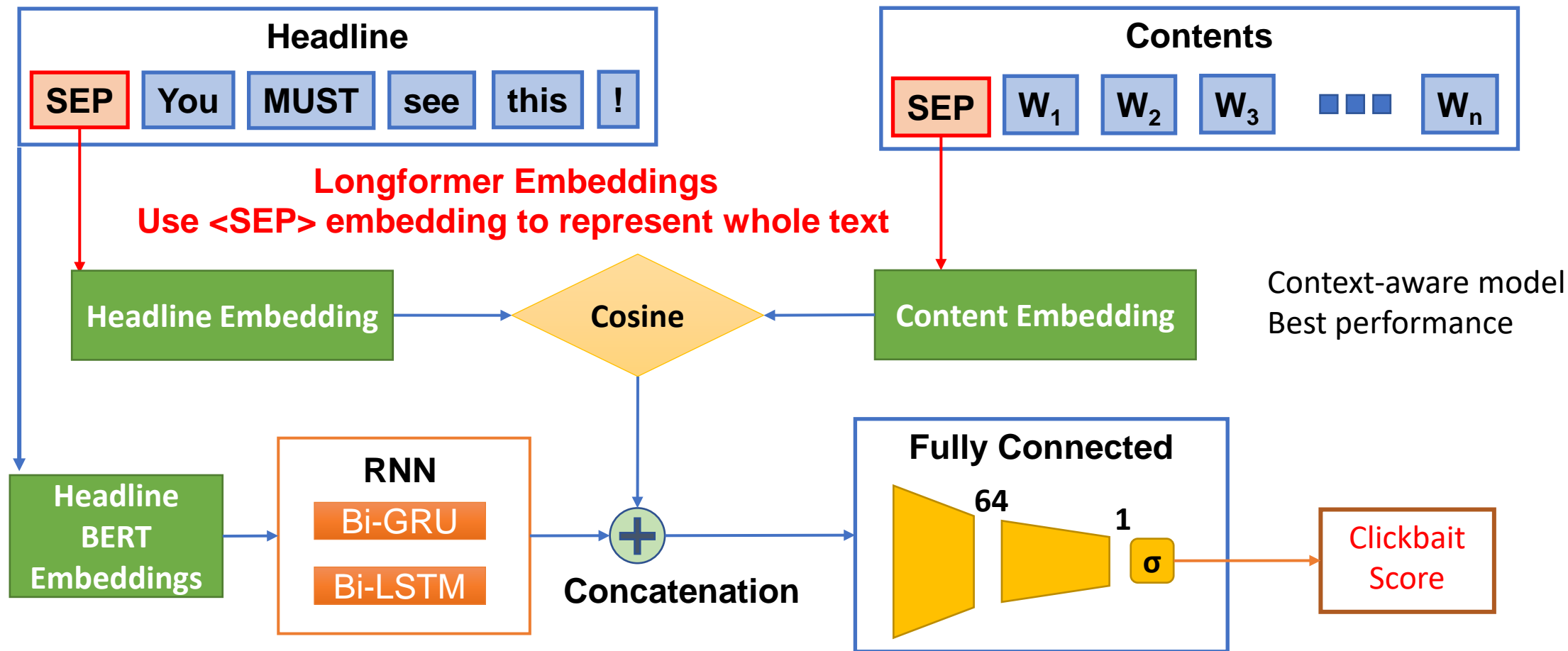


# Headline BERT with RNN Network





# BERT + Longformer with Parallel Structure



# Experiments - Training/Evaluation Details

- **Train/Validation Split:**

	#tweets	#clickbaits	#non-clickbaits
Training	11663	2027	9636
Validation	1300	203	1097
Total	12963	2230	10733

- **Evaluation Metrics:**

- Benchmark with previous models *on the Clickbait 2017 Challenge for Webis-17 dataset*
- Regression: Mean Square Error (MSE)
- Classification: Accuracy & F1-Score

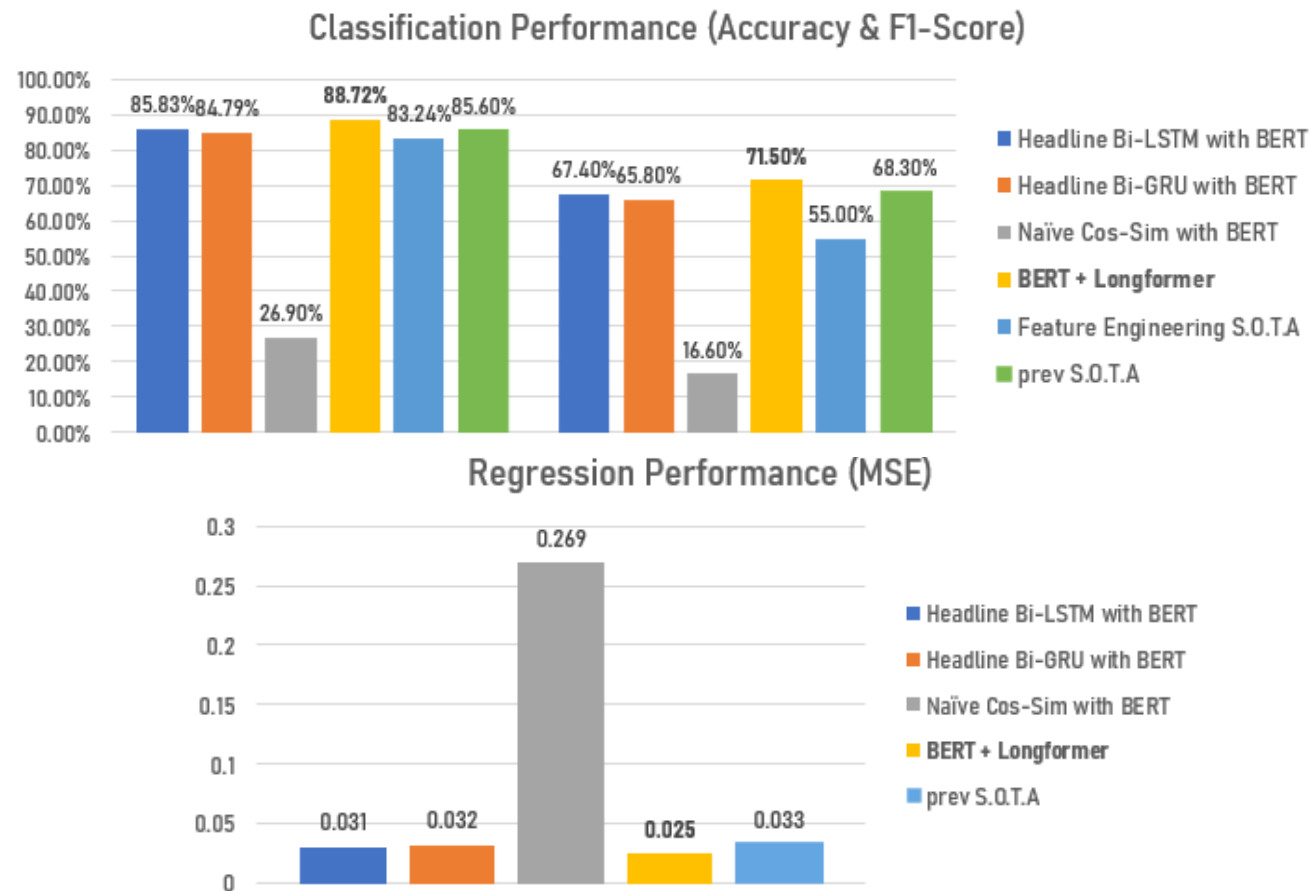
# Experiments - Training/Evaluation Details

- Hyperparameters for Bi-LSTM / Bi-GRU:
  - Hidden dimension: 50
  - # of layers: 2
  - Dropout: 0.2
- Hyperparameters for FC layer:
  - Hidden dimension: 64
  - Dropout: 0.2
- Loss function: MSE
- Optimizer: Adam
  - Learning rate:  $1e-4$
  - Adaptive Learning rate (Learning rate scheduler): decay by factor of 0.25 with patience of 2
  - Weight Decay:  $1e-3$
- Mini-batch size: 8 (limited RAM on Google Colab)
- # Epochs: 20
- Training time: Headline BERT – 0.5 day on T4, BERT + Longformer – 4.5 days on P100

# Experiments - Results

*Achieve S.O.T.A Performance on Webis-17!*

- **Surpass previous S.O.T.A**
  - ~3% w.r.t classification metrics
  - ~25% w.r.t regression MSE
- **Baseline 1 (naïve cos-sim) failed**
  - Average pooling across words failed to represent headline precisely
- **Baseline 2 (only headline) achieve comparable performance with previous S.O.T.A**



# Experiments – Case Study

## Success case - non-clickbait

**Headline:** "Tokyo's subway is **shut down** amid fears over an imminent North Korean **missile attack** on Japan"

**Content:** "One of Tokyo's major subways systems says it **shut down** all lines for 10 minutes after receiving warning of a North Korean **missile launch**. Tokyo Metro official Hiroshi Takizawa says the temporary suspension affected 13,000 passengers this morning. Service was halted on all nine lines at 6:07 am and was resumed at 6:17 am after it was clear there was no threat to Japan. Takizawa said it was the first time service had been stopped in response to a missile launch.... "

**Truth Score:** 0 (non-clickbait)

**Predict Score:** 0.074 (non-clickbait)

## Success case - clickbait

**Headline:** "26 pictures guaranteed to make you laugh every time"

**Content:** "Just trust me. We asked the **BuzzFeed Community** to send us the funniest pictures on the internet. Want to be featured in similar BuzzFeed posts? **Follow the BuzzFeed Community on Facebook and Twitter!** BuzzFeed Home © 2017..... "

**Truth Score:** 1.0 (clickbait)

**Predict Score:** 0.895 (clickbait)

# Experiments – Case Study

## Failure case - idiom

**Headline:** "CenturyLinkVoice: **New product launch: Testing the waters with social media**"

**Content:** "Back in the day, companies assembled focus groups to **gather feedback on product prototypes**. Changes were then made based on this group's advice. Today, the same kinds of opinions are being collected **through social media**, making prelaunch research vastly more efficient and cost-effective. Trusted customers who offer their frank opinions often become valuable promoters of a product before and after launch informing their social media followers at no cost to its maker..."

**Truth Score:** 0.13 (non-clickbait)

**Predict Score:** 0.674 (clickbait)

## Failure case - human label error

**Headline:** "18 **uplifting documentaries** guaranteed to put a smile on your face"

**Content:** "The real world isn't all trash. We asked the BuzzFeed Community to tell us their **favourite uplifting documentaries**. Here are the results.  
1. Twinsters (2015) "It's an uplifting story of two twins finding each other after they'd been adopted to families in different countries. An easy watch, and so heartwarming!" Watch on: Netflix Worldwide  
2. Iris (2014) "I cannot speak highly enough of this film..."

**Truth Score:** 0.73 (clickbait)

**Predict Score:** 0.221 (non-clickbait)

# Conclusion & Future Work

In this project, we proposed **Click-BERT: Clickbait Detector with Bidirectional Encoder Representations from Transformers**:

- Achieved S.O.T.A performance on the Webis-17 dataset
- Able to distinguish clickbait vs. non-clickbait contents with high accuracy

## Future Work:

- Directly fine-tune BERT-like model (compute limits)
- Improve language understanding abilities for media idioms
- Incorporate feature engineering

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**Thanks for listening!**  
**Any Questions?**

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