# You Are What You Annotate: Towards Better Models through Annotator Representations

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# Inherent annotation disagreements

#### Friends QIA

Question: Did Rachel tell you we hired a male nanny?

Answer: I think that's great!

ANN Answer (1), Not the Answer (2), Answer Subject to Some Conditions (3), Neither (4), Other (5): 1, 1, 4

### **Pejorative**

Text: @WORSTRAPLYRICS Everything Jay-Z writes is trash.

ANN PEJORATIVE (1)  $\leftarrow$  Non-Pejorative (0): 1, 0, 0

#### **HS-Brexit**

*Text:* RT <user>: Islam has no place in Europe #Brexit. ANN NO HATE (1)  $\leftarrow$  HATE (0): 1, 1, 1, 0, 0, 0

#### **MultiDomain Agreement**

*Text:* Please lost you yelling insanely at the sky on Nov 3 losers

ANN OFFENSIVE (1)  $\leftarrow$  NOT OFFENSIVE (0): 1, 1, 1, 0, 0

#### **Go Emotions**

Text: This is how I feel when I use a crosswalk on a busy street

ANN Positive (1), Neutral (0), Ambiguous (-1), Negative (-2): 1, 0

### Humor

Text A: Being crushed by large objects can be very depressing.

Text B: As you make your bed, so you will sleep on it. ANN WHICH IS FUNNIER, X MEANS A TIE: A, A, B, X, X

#### **CommitmentBank**

*Premise:* Meg realized she'd been a complete fool. She could have said it differently. If she'd said Carolyn had borrowed a book from Clare and wanted to return it they'd have given her the address.

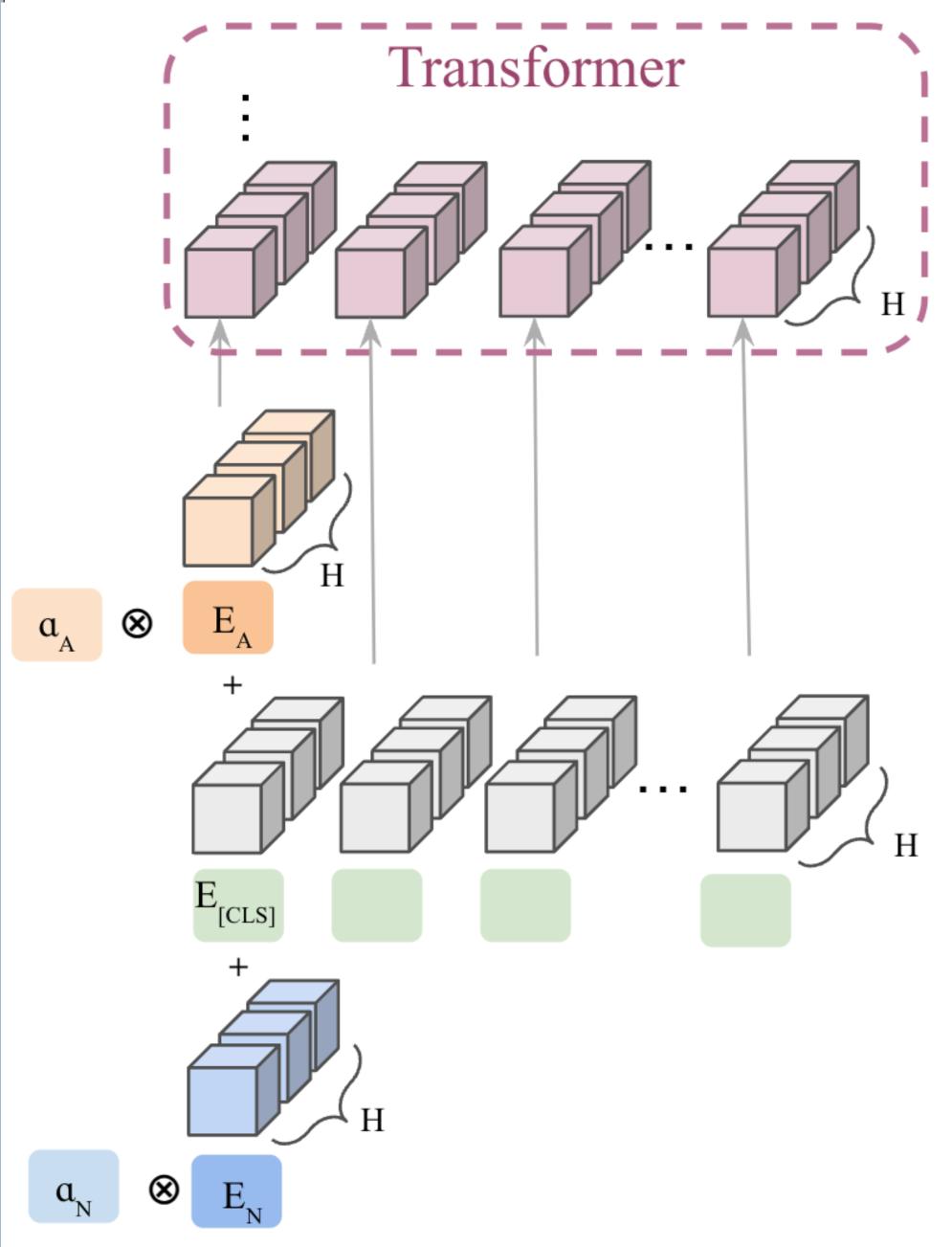
Hypothesis: Carolyn had borrowed a book from Clare. ANN ENTAIL (3) <->CONTRADICT (-3): 3, 3, 3, 2, 0, -3, -3, -3

#### **Sentiment Analysis**

*Text:* Even hotel bar food is good in California...fresh avocados, old chicken, and reasonably recent greens. Mmmm. Really.

ANN Positive (2) <->Negative (-2): 2, 2, 0, -1

Our Approach (Continued)



# Annotator-based predictions

*Text:* We know it anecdotally from readers we've heard from who've been blatantly discriminated against because they're older.

Positive  $(2) \leftarrow Negative (-2)$ 

<b>Annotator ID</b>	1	2	3	4
Gold	-1	0	-2	-2
T	-1	-1	-1	-1
$\mathbf{E}_n + \mathbf{E}_a$	-1	0	-1	-2

# **Hate speech Detection**

Problematic to ignore such disagreement!

aggregating labels → ignores the under-represented groups

## **Humor and Sentiment**

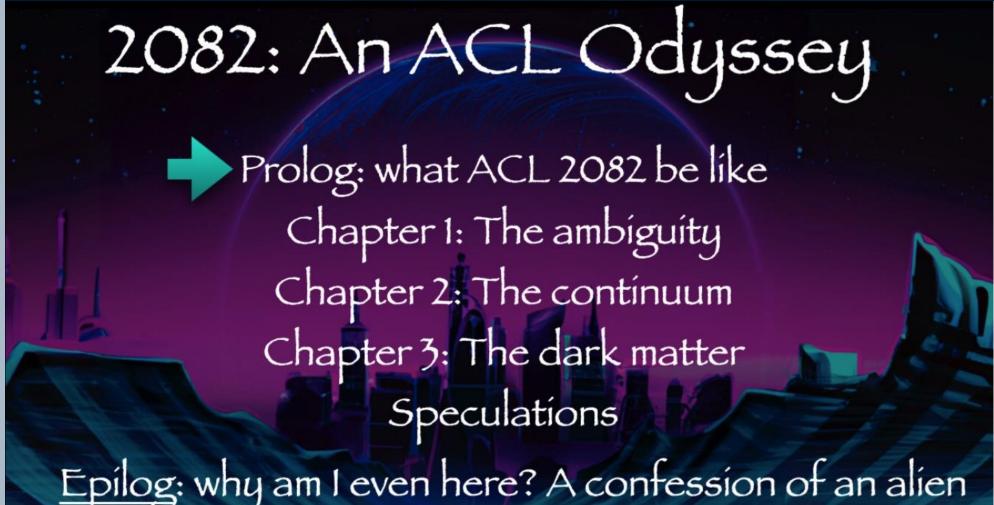
highly subjective

## Natural Language Inference (NLI)

Previous studies showed the inherent annotation disagreements

Instead, Let models learn from data that has inherent disagreement!

# Factors that cause annotation disagreements



Categories do exist, but the boundaries are "squish".

— Yejin Choi (University of

Washington)

Babara Plank's survey:

- Differences in interpretation
- Certain preferences
- Difficult cases or multiple plausible answers



Is the dress white and gold or black and blue?

Qualia

# Our Approach

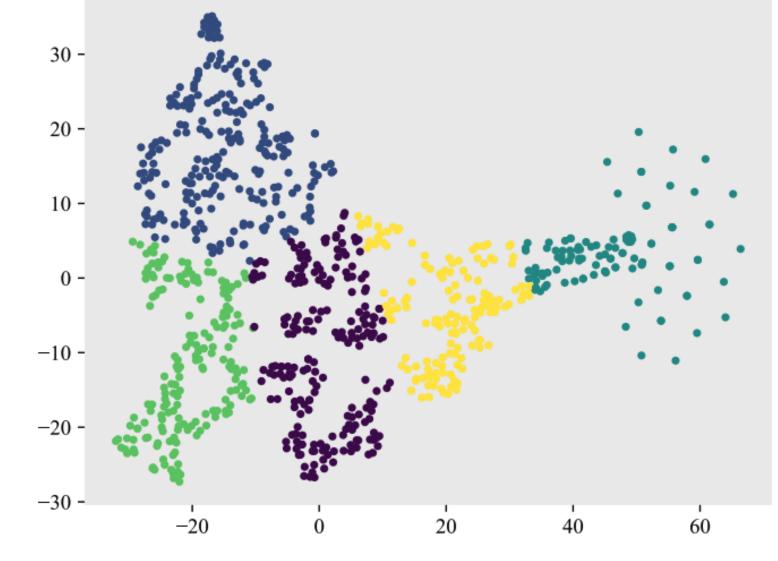
## Two representations:

- Annotator Embedding (E<sub>A</sub>): represent each annotator
- Annotation Embeddings (E<sub>1</sub>): aggregate annotators' annotations on other examples

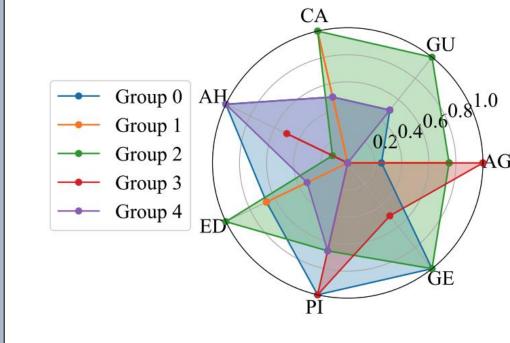
# Two weights:

balance the effects of text and the embeddings

# Grounded to real-world demographic features



# (a) Annotation Embedding



CA: current living area GU: grew up area

AG: age

GE: gender

PI: political identification

ED: education

AH: annual household income