

# Letting Emotions Flow: Success Prediction by Modeling the Flow of Emotions in Books

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## Motivation

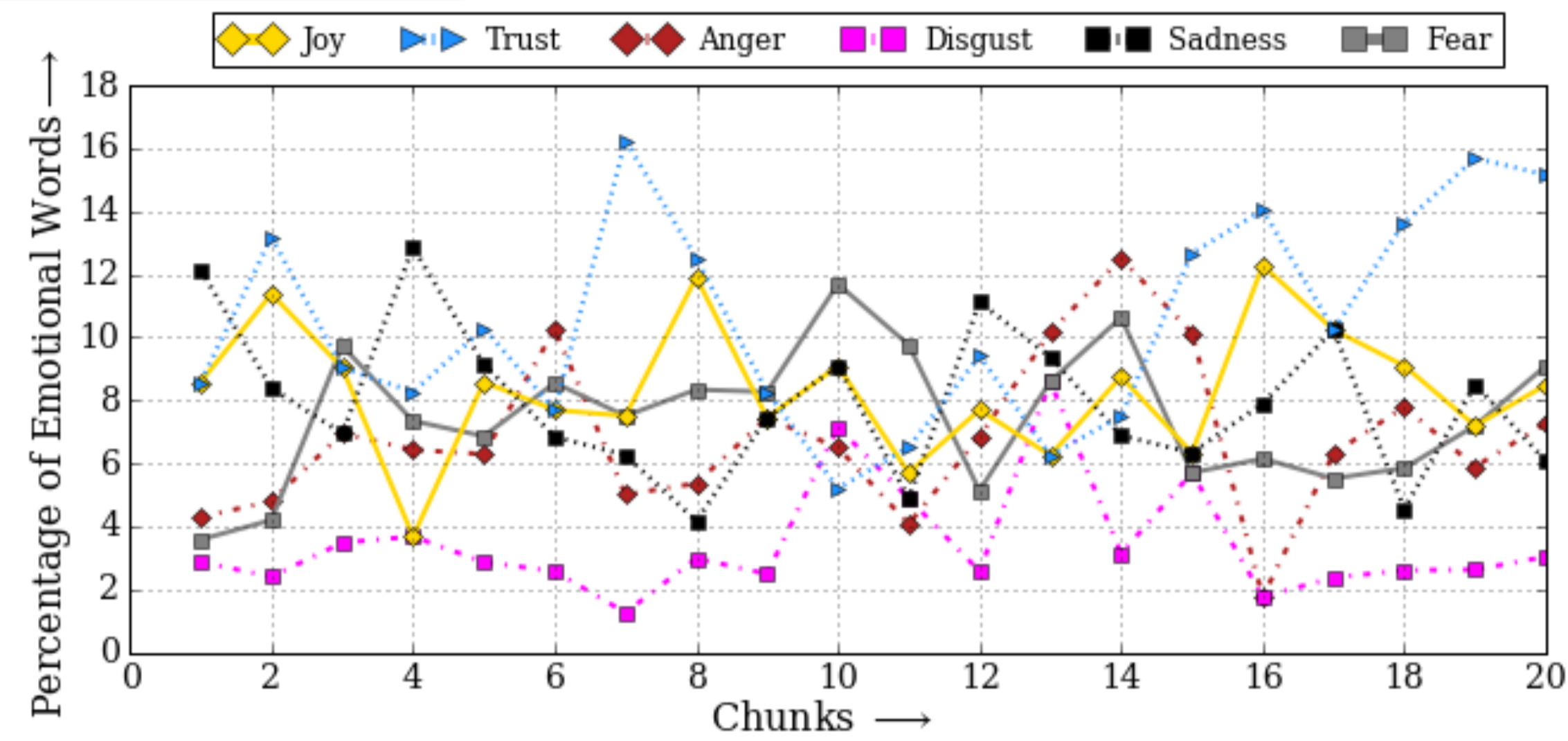


Fig 1: Flow of emotions in *Alice in Wonderland*

❖ Capture author's dexterity in use of emotions across book to predict book's likability

## Dataset

$$Success(x = Book) = \begin{cases} 1 & \text{if average\_rating}(x) \geq 3.5 \text{ and } \# \text{ rating}(x) \geq 10 \\ 0 & \text{if average\_rating}(x) < 3.5 \text{ and } \# \text{ rating}(x) \geq 10 \end{cases}$$

Category	Books
Successful	654
Unsuccessful	349

Table 1: Data distribution



## Multitask Emotion Flow Model

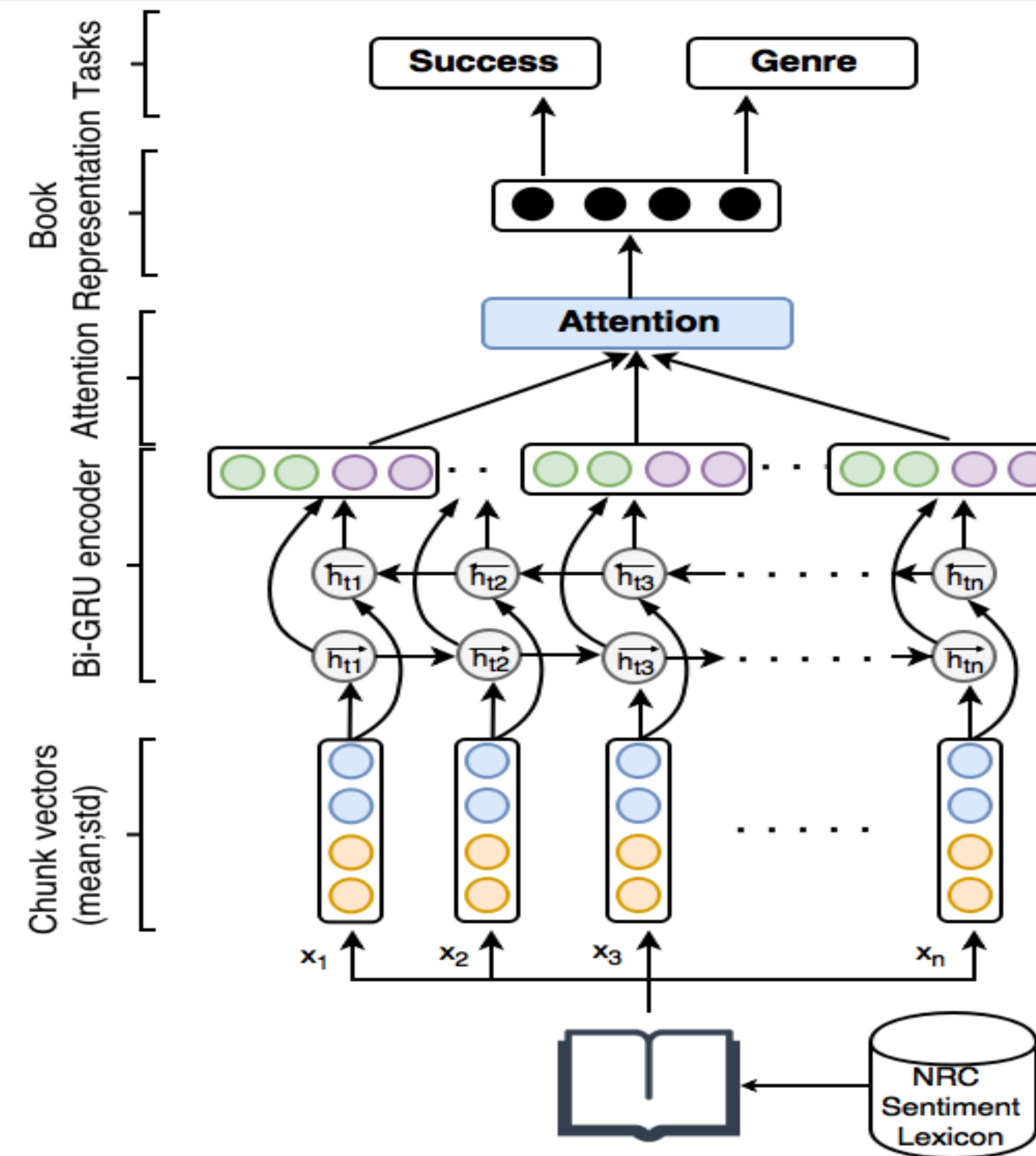


Fig 2: Multitask Emotion Flow Model

## Analysis

### Attention

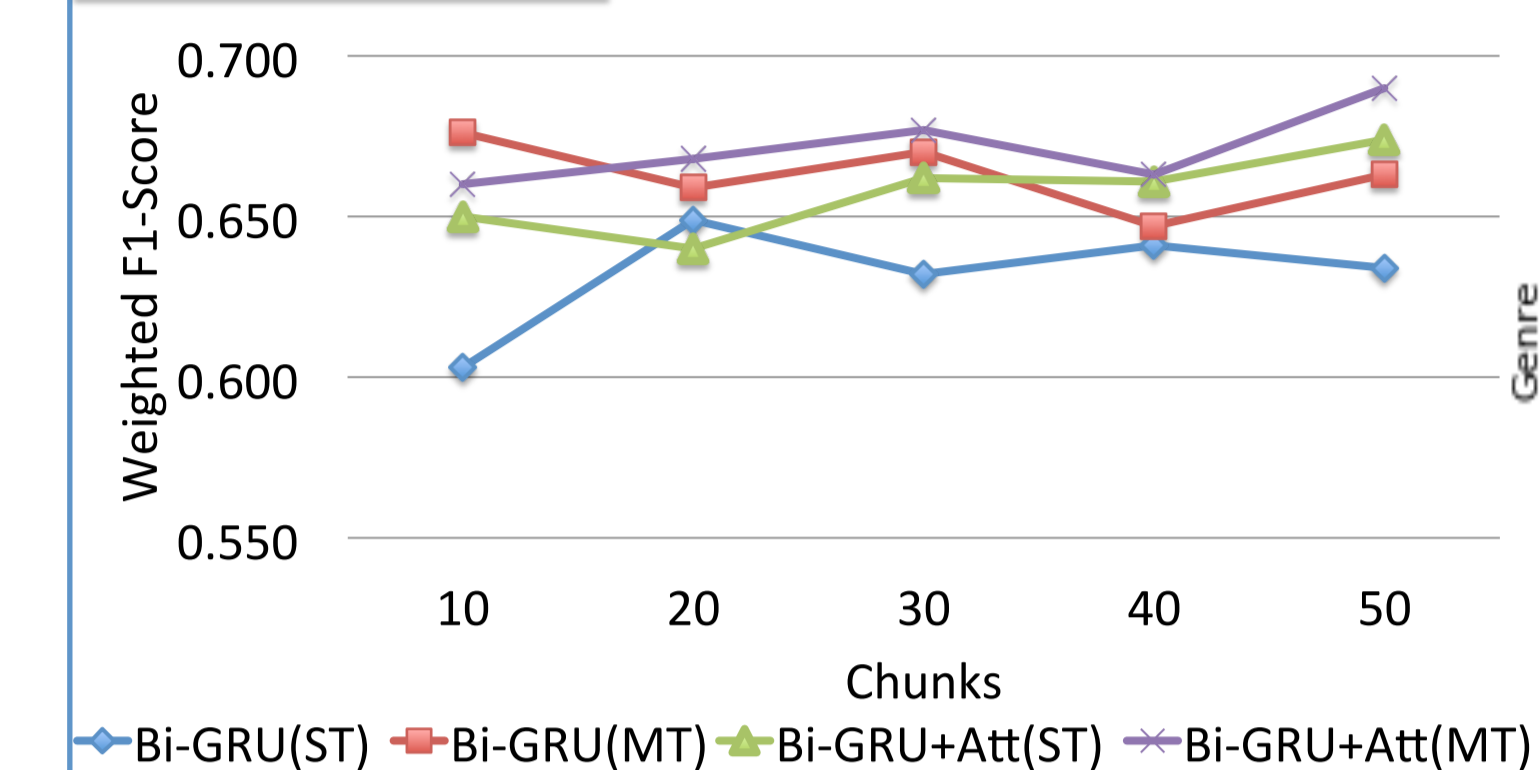


Fig 3: Comparison of the Emotion Flow with and without attention mechanism

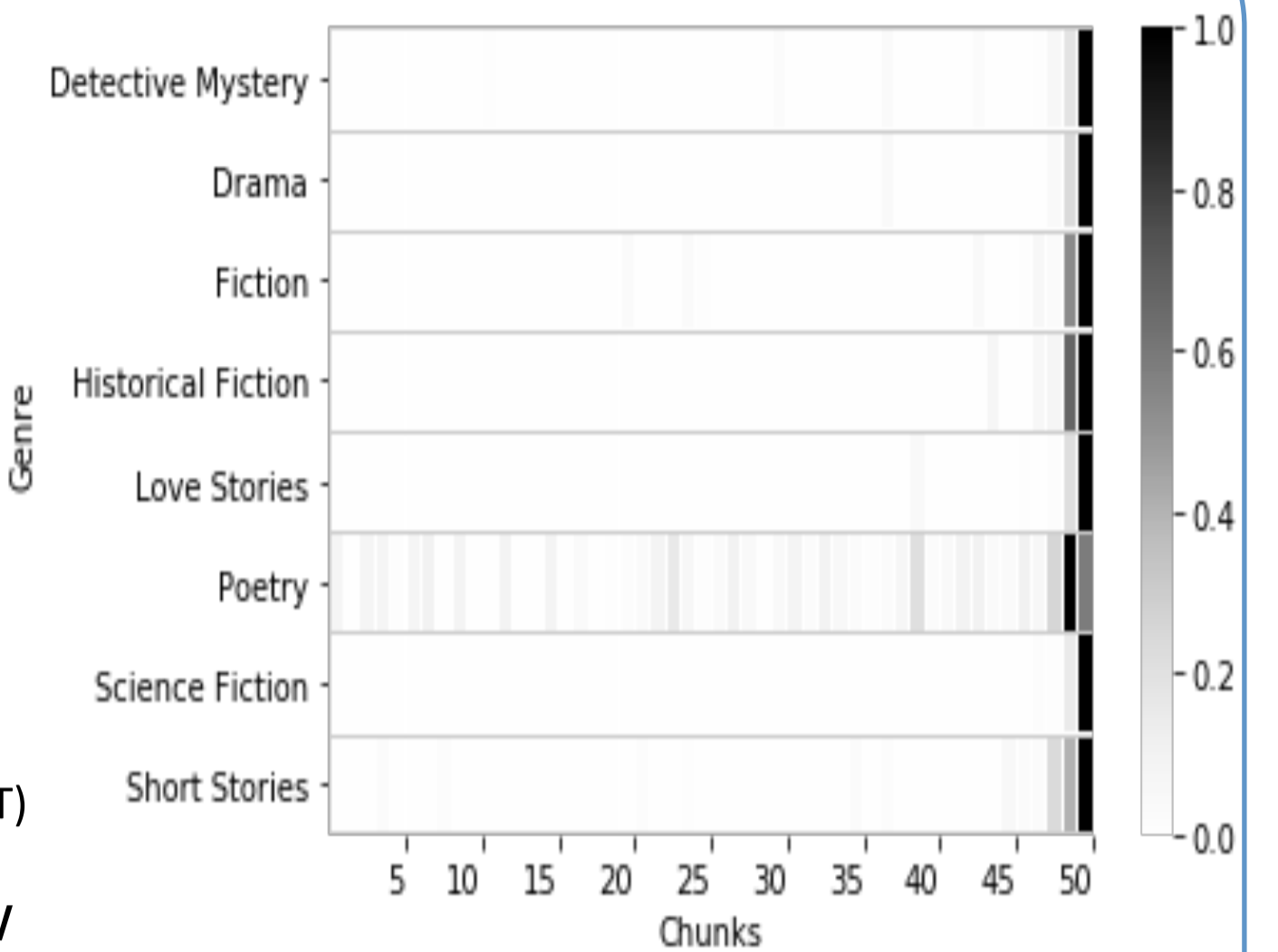


Fig 4: Mean attention weights per chunk per genre visualization

### Emotion Shapes

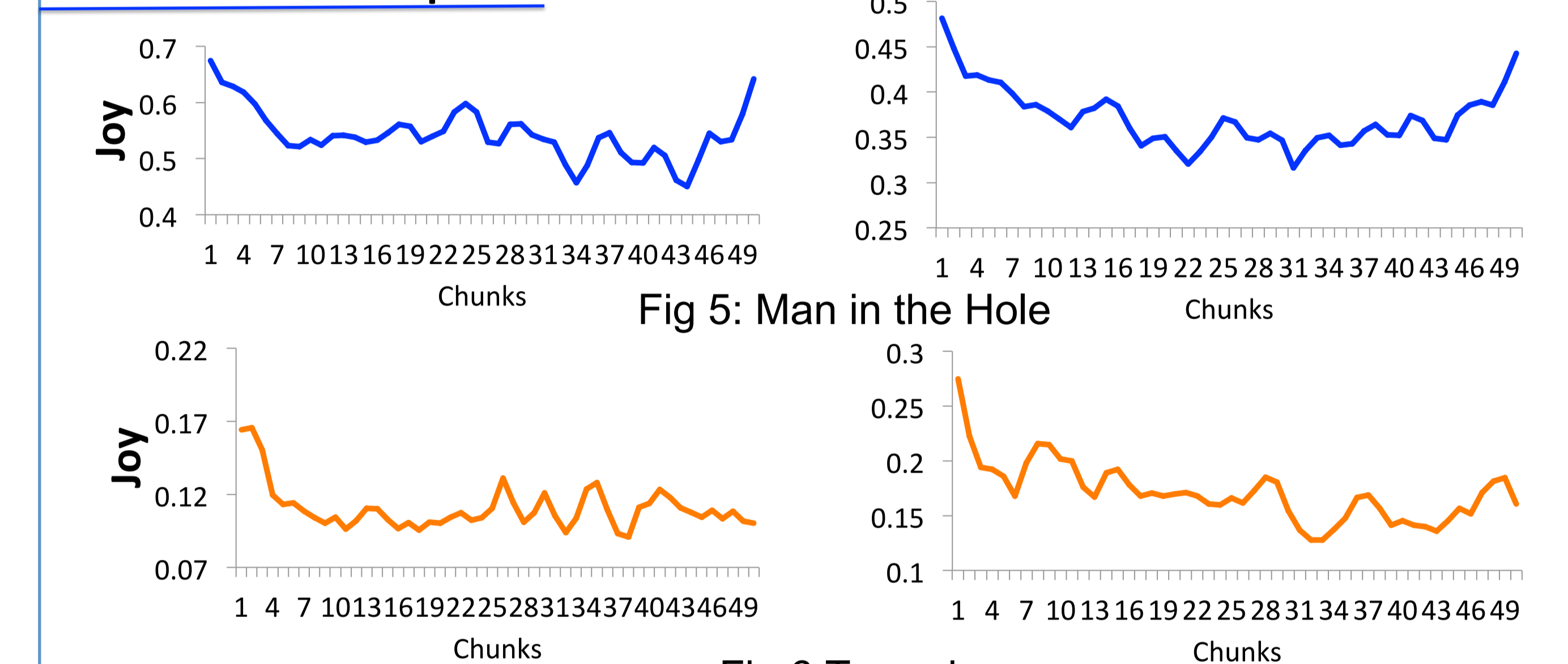
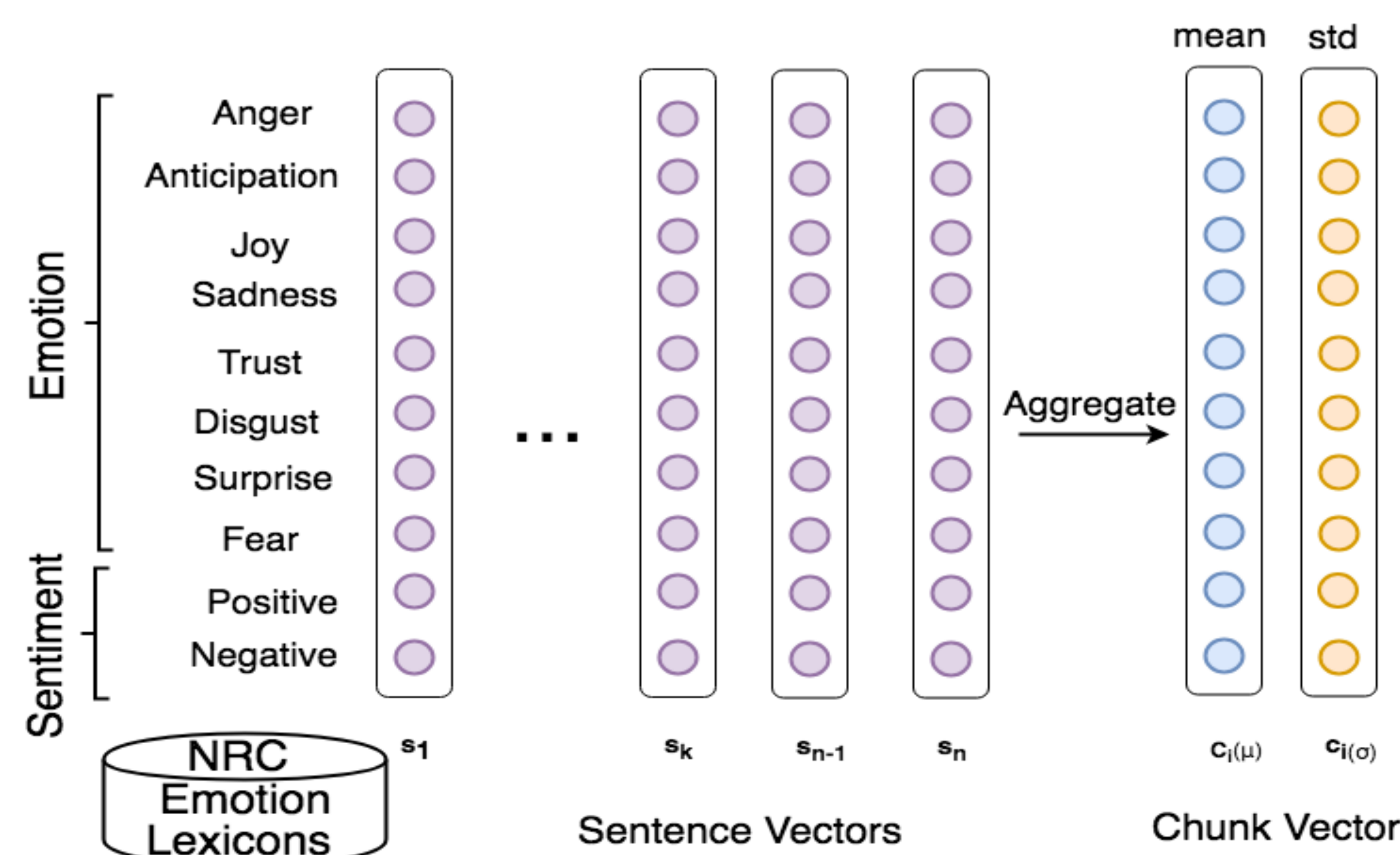


Fig 5: Man in the Hole

Fig 6: Tragedy

## Representation of Emotions



## Results

Book Content	Methods	Chunks	1000 Sentences		All	
			ST	MT	ST	MT
Majority Class		-	0.506	0.506	0.506	0.506
SentiWordNet + SVM		20	0.582	0.610	-	-
NRC + SVM		10	0.526	0.597	0.541	0.641
NRC + SVM		20	0.537	0.590	0.577	0.604
NRC + SVM		30	0.587	0.576	0.595	0.600
NRC + SVM		50	0.611	0.586	0.597	0.636
Emotion Flow		10	0.632	0.643	0.650	0.660
Emotion Flow		20	0.612	0.639	0.640	0.668
Emotion Flow		30	0.630	0.657	0.662	0.677
Emotion Flow		50	0.656	0.666	0.674	<b>0.690</b>

Table 2: Weighted F1-scores for success classification

### Climax Emotions

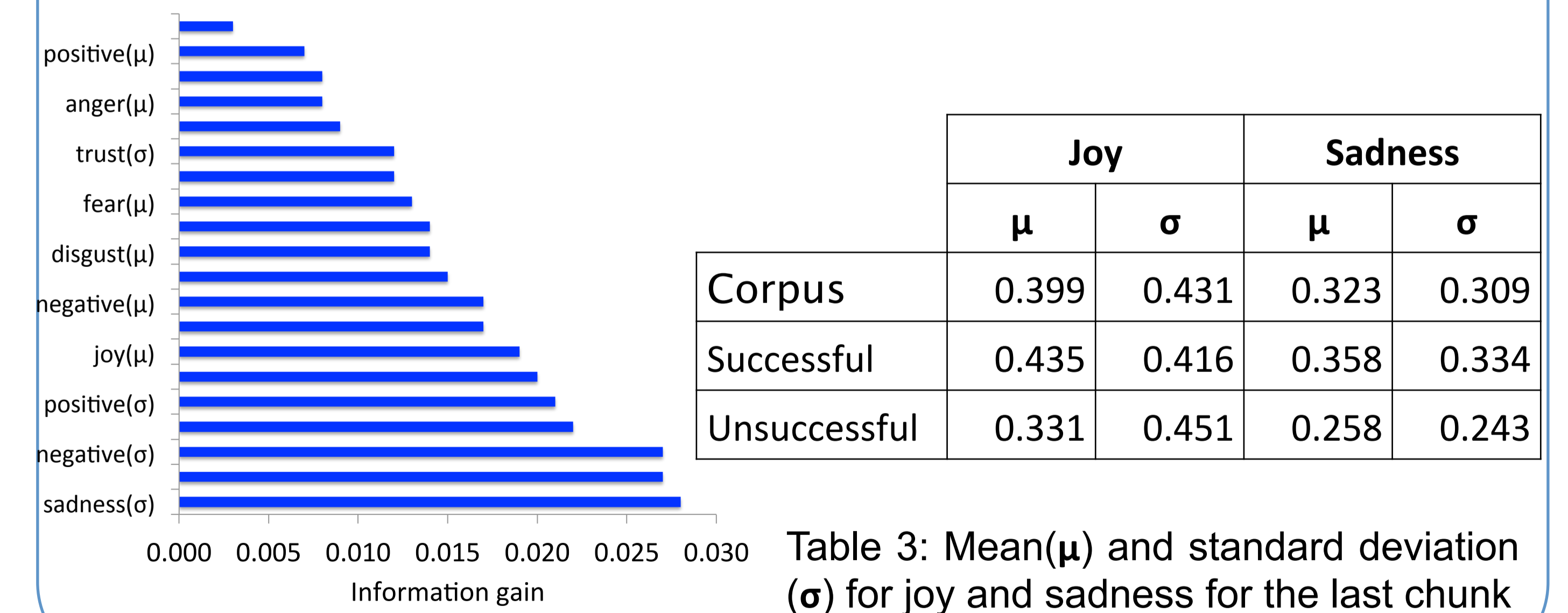


Fig 7: Feature ranking

Corpus	Joy		Sadness	
	$\mu$	$\sigma$	$\mu$	$\sigma$
Corpus	0.399	0.431	0.323	0.309
Successful	0.435	0.416	0.358	0.334
Unsuccessful	0.331	0.451	0.258	0.243

Table 3: Mean ( $\mu$ ) and standard deviation ( $\sigma$ ) for joy and sadness for the last chunk

## Conclusions and Future Work

- ❖ Modeling emotions as a sequential flow improves prediction accuracy
- ❖ Most attention weights are given to the last chunk for all genres except for *Poetry*
- ❖ Readers enjoy endings with stronger/well-defined emotions
- ❖ Use hierarchical methods with logical grouping to model emotion flow



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NAACL  
June, 2018  
New Orleans, Louisiana