

Clash of Crowds: Understanding Human Text Classification through Crowdsourcing Games

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INTRODUCTION

In order to design interactive machine learning systems that better coordinate human and computer, system designers need to have a good understanding on how human and computer make their decisions. For example, in order to design an interactive machine learning system for text sentiment classification, one needs to know which words in the text are features used by humans and computers to make their decisions (whether a document is positive or negative). While features used by machine learning algorithms, e.g., n-gram model, are well understood, little is known about what are the text features used by human in their decision making process. Our project seeks to understand the human labeling process (the connection between labels created by human and feature words influence them most in the labeling process) in the context of *text sentiment labeling tasks*.

Prior work seeks to understand features that human used in labeling text sentiment by mapping users' mental model into concrete words in the document; the approach used to collect such feature words (i.e., "human features") is to directly ask users to (1) label a document and (2) then select words associated with this decision. While such techniques are straightforward and easy to collect, it is unclear whether such features suffer from the "hindsight bias" effect of human labelers: post-hoc word selection could lead users to "make up" reasons (i.e., words) that they believe best explain their completed labeling decision, rather than the ones that reflect their instincts. If such effect exists, collected labels and features are not guaranteed to be naturally paired.

In order to collect more meaningful features and provide better explanation of these features, our key insight is to collect features along with labels during the labeling procedure through interactive games. Concretely, we designed a set of 5 crowd sourcing games: *DirectAnnotation*, *Bingo*, *HangmanAdd*, *HangmanSub* and *Censor*. While these games share the same goal, players in different games have different ways to access information in target documents. These controlled

variance in the game lets us both collect and verify features and labels for every target document, as we will present later in Section ???. Though our implementation and deployment of the game, we observed the following set of facts that can potentially be used to guide interactive machine learning system design.

The rest of the report structured as follows: (1) the text classification task we aim to solve (Section 2), (2) our game designs, (3) our deployment of the games and evaluation.

TASK SELECTION

The text classification task used in our paper is sentiment labeling for IMDB movie reviews. Labeling sentiment of movie reviews is a challenge task since (1) many reviews contain humorous or ironic elements that makes the movie sentiment orientation tricky to identify, and (2) movie reviews are self-labeled: sentiment labeling of them are directly posted along with the review, which makes the training data and ground truth easy to obtain.

GAME DESIGN

We have designed the following five games to collect text sentiment labels and features from players; their interfaces are shown in Figure 1.

- *DirectAnnotate*. This is the base game where the player directly labels document sentiment and selects words that influence decision making (informative words).
- *Bingo*. This is the game where two players coop on labeling document sentiment and extract informative words. The game requires labels collected from the team are the same, and the game is finished when two players reach an agreement on four informative words.
- *HangmanAdd*. Players in this game are assigned as annotator and guesser. The annotator labels the document and provides a list of ranked informative words to assist the guesser to guess the document sentiment. Game result counts on how many words are used and whether the guesser result agrees with the annotator's label.
- *HangmanSubtract*. This is a competitive game where the two players are assigned as roles masker and guesser. The masker is asked to label the document and mask informative words that are important to his decision. The masker's goal

add
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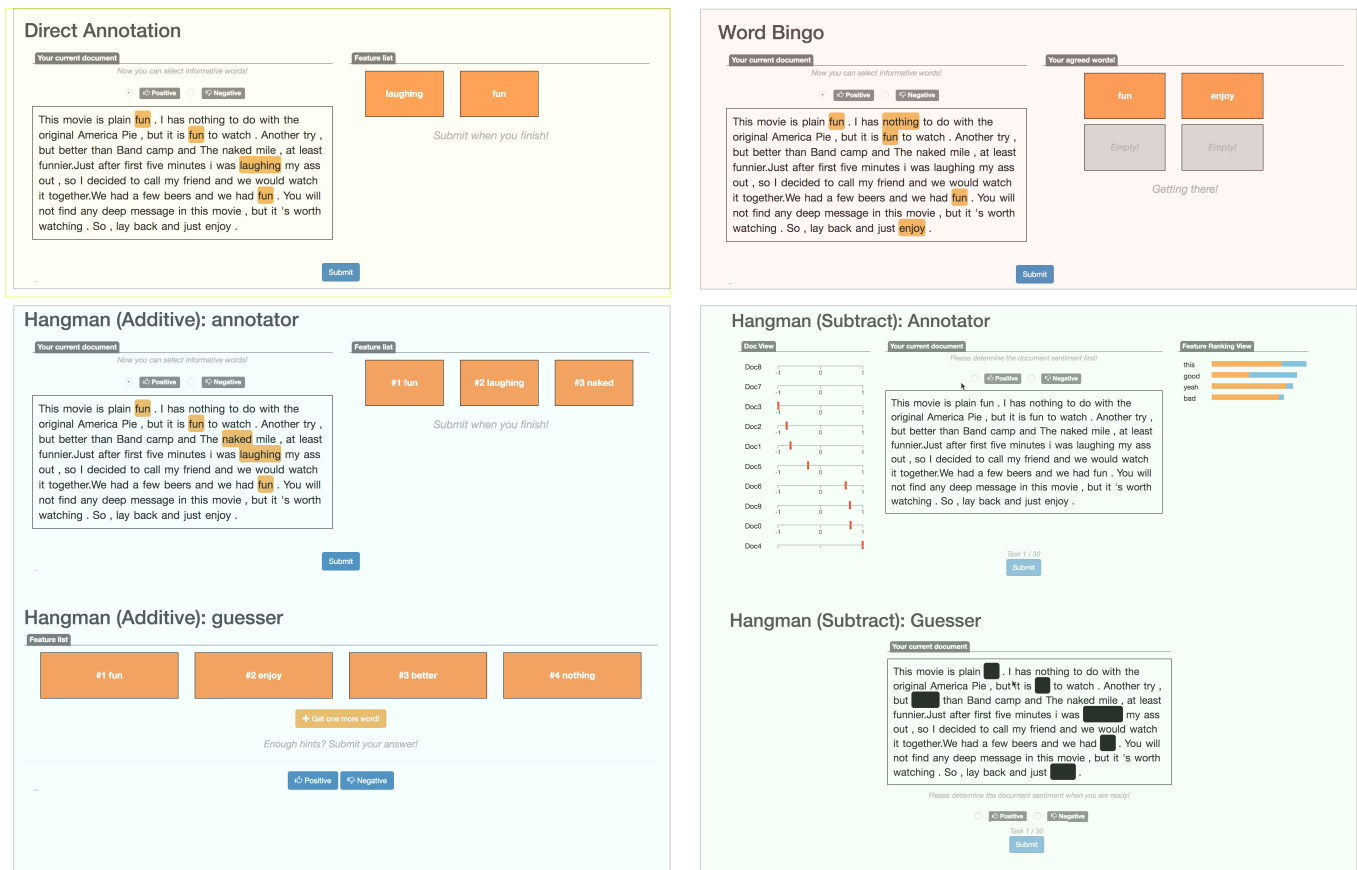


Figure 1. Interface of DirectAnnotate, Bingo, HangmanAdd and HangmanSubtract

is to hinder the guess to figure out the correct the document sentiment and the guesser is asked to guess document based on the masked document. (The guesser can unmask words with the reduction on the payment bonus and the masker earn more if labeled less but more effective.)

• Censor.

Our games are designed based the three criteria below, and our game design details are presented in Table 1.

- **Data quality:** Players of the game should be able to collect less noisy or biased labels and features through the game, and preferably, collected features can be verified.
- **Difficulty:** The game should be fairly intuitive to play and should not overwhelm the players.
- **Compatible incentive:** The incentives for different players, either in competitive or collaborative games, should be compatible such that neither sides have shortcuts and could play in a fair setting.

IMPLEMENTATION & DEPLOYMENT

Our games are deployed through....

DATA ANALYSIS

Collected Features

The set of features that can be collected from different games are presented below.

Data	Ann.	Bin.	H.Add	H.Sub
Labels-full	•	•	•	•
Labels-verify		○	•	•
Features-init	•	•	•	•
Features-rank		•	•	•
Features-refine				•
Features-verify		•	•	•

Labels-full refers to whether we can extract sentiment label from the game, *labels-verify* refers to whether the game supports different players to verify their label result, *feature-init* refers to whether the game obtains features provided by the gamer, *feature-rank* refers to whether obtained features are ranked, *feature-refine* refers to whether the player can refine initially collected feature, and *feature-verify* refers to whether features are verified through multiple players.

CONCLUSION

FUTURE WORK

REFERENCES

add its content

add details

GameSystem	Player	Mechanism	Pros&Cons	Compatible Incentive
Direct Annotation (Baseline)	1P	Directly ask users to choose features they used.	<ul style="list-style-type: none"> + Access to the full doc - * Hindsight bias - * No checks 	N/A (not a game)
Bingo	2P, Collab	Two users annotate one document. Score when they label the same feature.	<ul style="list-style-type: none"> + * Less noisy + Access to the full doc - * Hindsight bias - * Low representation of individual users 	Reward Mutual agreement
Hangman, <i>Additive</i>	2P, Collab	Guesser asks for a word from annotator. Annotator returns a word to the guesser.	<ul style="list-style-type: none"> + * Mediate hindsight bias + Engaging - Takes more time - Guesser cannot access document structure 	<ul style="list-style-type: none"> - Reward agreement - Penalize queries used
Censor	2P, Compete	Censors mask words in a document. Identifiers label the censored document.	<ul style="list-style-type: none"> + * Keep doc structure + * Mediate hindsight bias + Engaging and easy - Can't reuse identifiers - Need to retain censors 	Censors: Reward fooling users Identifiers: Reward for being correct
Hangman, <i>Subtractive</i>	2P, Compete	Censors remove a set of words from a document. Guesser queries one word at a time (can choose position).	<ul style="list-style-type: none"> + * Keep doc structure + * Mediate hindsight bias + Engaging and easy - Can't reuse guessers - More mental load balancing prices 	Censors: Reward fooling users Identifiers: Reward for being correct

Table 1. Designed games. Pros and Cons with “*” are the attributes affecting data quality. In the table, *Game Mode* indicates the number of the players in the game and whether it is collaborative or competitive, *Mechanism* presents how labels and features are obtained from the player through the game, *Pros&Cons* shows what is the game designs strength and weakness and *Incentive Mechanism* indicates what makes the game incentive to the players.