## **CMPT 318 Project Results**

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#### **ACM Reference Format:**

### 1 HYPERPARAMETER STUDY

A grid search of 5 hyperparameters was conducted.

- selection Use sentiment readings from either the beginning, middle, or end of the tokenized steam reviews. Has values ('left', 'middle', 'right') in the code, respectiely
- k number of columns to use in SelKBest(). (5, 2) were checked for k as well as using all of the 10 relevant columns.
- C, gamma, and kernel These are used directly as parameters to the grid search of SVC. Values (1, 10, 0.1), ('scale', 'auto' and ('linear', 'rbf', 'sigmoid') were checked, respectively.

#### 2 MAIN RESULTS

With the precision varying only between 0.87 and 0.88, I cannot say that I achieved a high precision, but I can discuss what parameters affected results in the grid search the most. As far as selection, selecting the rightmost tokens performed slightly better than middle on average, and left performed significantly worse. The case was similar for kernel selection. rbf performed slightly better than linear, and sigmoid much worse than both. The C parameter seemed to increase potential performance slightly linearly

for my chosen values of 0.1, 1, and 10. It makes sense that a larger C would allow better performance, if the correlation between sentiments and ['user\_suggestion'] is smooth and not noisy. Finally, the k in SelKBest did not affect performance much. This was to be expected since the non-sentiment columns are hardly even correlated with [user\_suggestion]. In the end, the most accurate and precise classifiers used an rbf kernel and sentiment readings from the final 512 tokens.

#### 3 ROADBLOCKS

There is one main roadblock worth mentioning. While running sentiment analysis, I discovered that the twitter transformer and most huggingface transformers only support 512-token inputs, too small for roughly half of the reviews. After some research, I found out that selected only a substring of tokens is a valid solution and that selecting the beginning, middle, or end can have different results. Supposedly for long paragraphs or essay-like documents, selecting the end tokens yields more accurate sentiment readings since that is where the concluding remarks are found. This seemed to prove true as is shown in the figures.

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Data #	columns (total 5 Column	columns): Non-Null Count	Dtype	Data #	columns (total 8	columns): Non-Null Count	Dtype
0	review_id	17494 non-null	int64	0	review_id	17316 non-null	int64
1	title	17494 non-null	object	1	title	17316 non-null	object
2	year	17316 non-null	float64	2	year	17316 non-null	float64
3	user_review	17494 non-null	object	3	user_review	17316 non-null	object
4	user_suggestion	17494 non-null	int64	4	user_suggestion	17316 non-null	int64
<pre>dtypes: float64(1), int64(2), object(2)</pre>				5	is_early_access	17316 non-null	bool
memory usage: 683.5+ KB				6	received_free	17316 non-null	bool
				7	contains_art	17316 non-null	bool
				<pre>dtypes: bool(3), float64(1), int64(2), object(2)</pre>			
				memory usage: 862.4+ KB			

# Original

## After Processing

Fig. 1. Dataset: The dataset was unchanged.

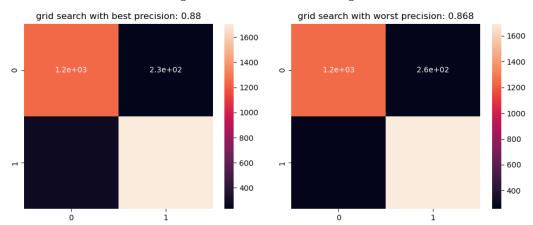


Fig. 2. Results: Confusion matrices of classifiers with the best and worst precisions



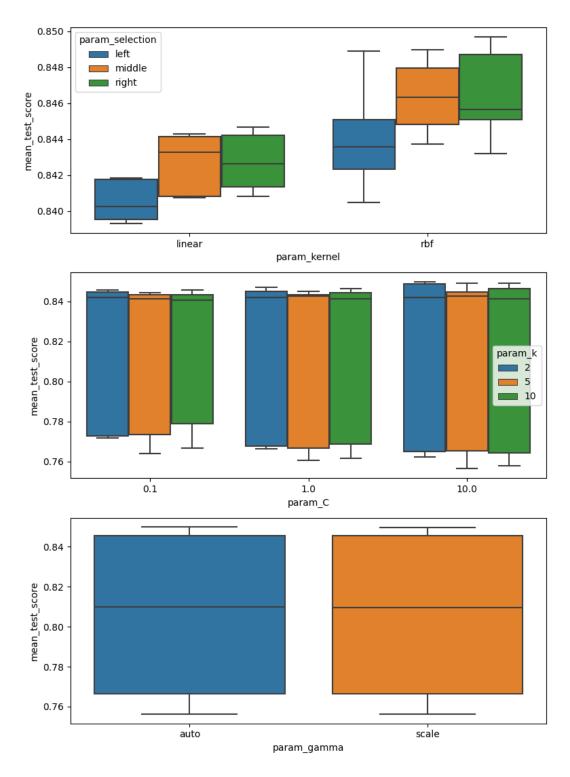


Fig. 3. Additional Results: Boxplots showing means of classifier accuracies accross the different hyperparameters