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**CSE 158** 

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## **Assignment #1 Report**

## Play Prediction Task

For this task, we were asked to predict whether a random given user would play a given game. This is a supervised classification problem because we are trying to classify whether the user would (1) or wouldn't (0) play the game. I used logistic regression to make a model that can accurately predict this given various features. When first exploring the training set, I noticed the distribution of 'hours\_transformed' was skewed right, and there were some really high values, and decided to remove all 'hours\_transformed' that was greater than 8.5, which ended up removing ~ 5% of our data, which is reasonable considering we have well over 150,000 data observations. Most importantly, all our training data consists of positive (1) classifications, and wouldn't make for a balanced model, therefore for every positive user, I created an observation of a game they didn't play, doubling my training set size. For my features I used Jaccard Similarity, Popularity threshold (totalPlayed/1.5) with average hours the user plays, average hours spent on the game, and how often the game is offered for free. For the Jaccard similarity, I did a user-user comparison by comparing how similar the users of the given game are with the users of all games played by the given user. In my feature vectors, I added whether the maximum Jaccard Similarity was greater than 0.021, optimizing BER.

## **Category Prediction Task**

For this task, we were asked to predict the 'genreID' based on the textual features of a given review. This is a supervised multi-classification problem because we are using textual features to classify the review to a specific genre of games. I also used logistic regression for this problem using a bag of words model to create my features. I used the 3500 most occurring words out of all reviews, making the matrix of size (nObs, 3501), including the constant. Additionally, I added the length of the review in numWords making each feature vector of length 3502. The max\_iter parameter had to be increased as we have lot of data, and our model performed better when our regularizer constant (C) was 10. I tried using the TF-IDF approach, by using ~3500 of the highest TF-IDF values in the dataset, however this reported a lower accuracy so I chose to use my bag of words model.