# MovieLens Capstone Report

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## Introduction

This capstone project is part of the HarvardX PH125.9x Capstone assignment. The goal is to develop a movie recommendation system using the MovieLens 10M dataset and evaluate its performance based on RMSE. This report outlines the methodology, analysis steps, modeling results, and final model performance.

## Methods / Analysis

### Data Acquisition and Preparation

#### Validation Partition from edx

```
set.seed(1, sample.kind = "Rounding")
## Warning in set.seed(1, sample.kind = "Rounding"): non-uniform 'Rounding'
## sampler used
```

```
index <- createDataPartition(y = edx$rating, times = 1, p = 0.1, list = FALSE)
train_set <- edx[-index, ]
temp <- edx[index, ]
validation <- temp %>%
    semi_join(train_set, by = "movieId") %>%
    semi_join(train_set, by = "userId")
train_set <- train_set %>%
    semi_join(validation, by = "movieId") %>%
    semi_join(validation, by = "userId")
```

#### **RMSE Function**

```
RMSE <- function(true_ratings, predicted_ratings) {
   sqrt(mean((true_ratings - predicted_ratings)^2))
}</pre>
```

#### Models Tested

#### 1. Naive Model

```
mu_hat <- mean(train_set$rating)
naive_rmse <- RMSE(validation$rating, mu_hat)
rmse_results <- tibble(method = "Naive Mean Model", RMSE = naive_rmse)</pre>
```

#### 2. Movie Effect Model

#### 3. Movie + User Effect Model

```
user_avgs <- train_set %>%
  left_join(movie_avgs, by = "movieId") %>%
  group_by(userId) %>%
  summarize(b_u = mean(rating - mu_hat - b_i))
predicted_ratings <- validation %>%
  left_join(movie_avgs, by = "movieId") %>%
```

#### Model Comparison Table

```
print(rmse_results)
```

## **Next Steps**

- Regularization of bias terms
- Matrix factorization (e.g., recosystem)
- Final model training on edx
- Test on final\_holdout\_test

### Results

Provide here the RMSE table, brief discussion of results so far, and planned improvements. Example:

The naive model yielded RMSE  $\sim 1.06$ . Adding movie effect improved RMSE to  $\sim 0.943$ . Including user effect further reduced RMSE to  $\sim 0.865$ . Further improvement will focus on regularization and matrix factorization.

## Conclusion

This project presented the development of a movie rating prediction algorithm. Several models were implemented and evaluated, leading to progressive reductions in RMSE. Future iterations will integrate regularization and matrix factorization to reach target RMSE < 0.86490.

## Final RMSE on Hold-Out Test Set

(To be filled only after selecting final model)

```
# final_rmse <- RMSE(final_holdout_test$rating, predicted_values_from_final_model)
# final_rmse</pre>
```

## References

- HarvardX PH125.9x Capstone Instructions
  tidyverse, caret, data.table documentation