# Puck Predictions: Unraveling the NHL Game Forecasting Riddle

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#### Abstract

The goal of this project is simple: predict the outcomes of NHL games from any given state. As simple as the problem statement is, however, the solution is not so straightforward. To solve this problem, we will use a variety of machine learning techniques, including logistic regression, XGBoost, and ARIMA models. Additionally, we utilize a form of MCMC to simulate the outcomes of games from any given state. Our hypothesis is that we will be able to successfully predict the outcomes of NHL games with a high degree of accuracy using these tools.

### 1 Problem Statement and Motivation

# 2 Data

Our data came from the hockeyR Github repository[1]. This repository contains an abundance of data about every NHL game that has occured since the 2010-11 season. This data includes information about the events that transpire in a game (hits, shots, goals, etc.), which teams are playing, who is on the ice, and the final score of the game. The data is stored in a series of .csv.gz files, allowing for easy access and manipulation.

Each game in a season is given a unique identifier (game\_id), which is constant across all events in a game. Every event that occurs in a game will be stored in the event\_type column. There are 17 unique event types, including things such as game start, faceoff, shot, hit, and goal. Most of these event types are not relevant to our analysis, so we remove them from the dataset. After removing the unnecessary events, we are left with nine

events: blocked shot, faceoff, giveaway, goal, hit, missed shot, penalty, shot, and takeaway. These events are attributed to the team and player that performs the event. We only take into consideration the team that performs the event and discard the player information.

The data also contains information about when the event occured. This appears in a variaty of formats, but we only use the <code>game\_time\_remaining</code> column. <code>game\_time\_remaining</code> starts at 3600 (60 minutes) and counts down to 0. If the game goes into extra time, i.e., it is tied after 60 minutes, <code>game\_time\_remaining</code> will be a negative value.

We found that our data did not contain any missing values that was not easily explainable. For example, if a game is starting, there will be no events for penalties, which will result in a NaN value in the penalties column. Additionally, any data that was confusing or not easily explainable (for example the home team having 7 players on the ice and the away team having 5), was manually verified by watching a clip of the game where the event occurred to make sure the event was recorded correctly. We did not find any incorrectly recorded events, so we did not remove any strange events from out dataset.

- 3 Methods
- 3.1 t-SNE, UMAP, and PCA
- 3.2 Regression and XGBoost
- 3.3 MCMC Game Simulation
- 3.4 Exponential Smoothing
- 4 Results
- 5 Analysis
- 6 Ethical Considerations
- 7 Conclusions

# References

[1] Dan Morse. hockeyR-data: A collection of hockey datasets for use with the hockeyR package. https://github.com/danmorse314/hockeyR-data, 2024. Accessed: March 2024.