# National Hockey League (NHL) Shots

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#### Introduction

- The NHL presents a unique opportunity for sports analytics, especially with the Stanley Cup occurring
- We were interested in analyzing a collection of shots from the 2021-22 season to find several ways to optimize team performance
- 160.573 shots analyzed with 22 data points



## **Research Questions**

- 1) What is the impact of <u>shot location</u> and <u>shot angle</u> on frequency of shots made, and how do these impact the likelihood of making a successful shot?
- 2) How do NHL teams differ in their <u>ability to convert shots</u> into goals, and how does this variation in goal conversion rates <u>correlate with shot accuracy</u> and effectiveness across different game periods?
- 3) What is the effect, if any, of the "home-field advantage" on scoring?

## Research Question 1

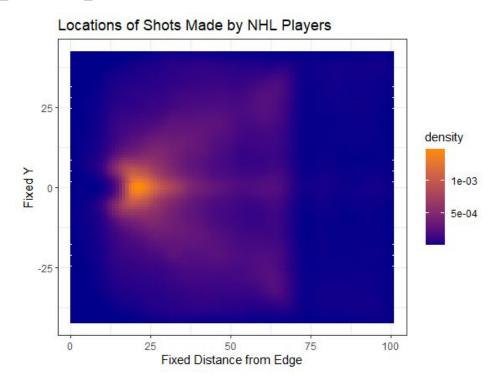
Where on the ice are goals more likely to be scored? How does the likelihood of scoring goals correlate with shot location and shot angle?

# **Shot Frequency by Location**

 Make new variable measuring the distance from the nearest edge to get insights for all teams in general rather than just home or away

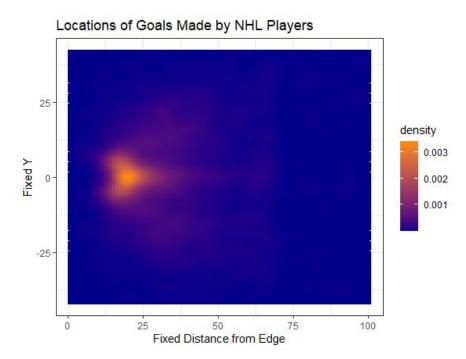
$$x_{\text{from\_edge}} = 100 - |x_{\text{fixed}}|$$

- Main insights:
  - Shots made most often close to the net
  - Noticeable frequency directly in front of the goal or at a 30-degree angle
  - Shots seem to cut off at ~70 units away from the edge
- Gives us a good sense of typical player behavior, may be result of training with best practices
- Preliminary data exploration before we approach our questions



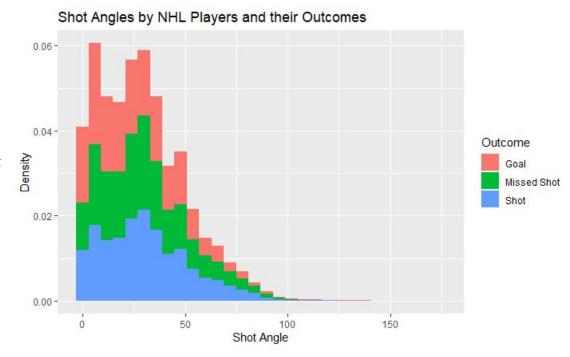
# Goal Frequency by Location

- Same premise as before, but only for goals
- Main insights:
  - Exhibits similar patterns to shot frequency
    - Most goals made near the net
    - Some goals made in front or 30 degrees away from the net
  - The cutoff happens at around 40 units
- Also gives a sense of player behavior and insights into best practices



### **Shot Angle Frequency and Outcomes**

- Gives a good sense of the marginal distribution of shot angles and outcomes, as well as conditional probability of outcome given shot angle
- Main insights:
  - Most shots tend to be "straight shots" (<50 degrees)</li>
  - Conditional distribution of outcome given angle about constant
  - More goals made with straighter shots



#### Regression Analysis: Goal Likelihood vs. Shot Location and Angle

Logistic regression model with the following formula:

```
goal \sim x\_{\rm from\_edge} + abs(y\_{\rm fixed}) + shot\_angle
```

- Assume "x\_from\_edge" is analogous to horizontal distance from the net (Very low frequency of shots made behind it)
- Main insights:
  - Statistically significant increases in goal likelihood with:
    - Shots made <u>horizontally and</u>
       <u>vertically closer</u> to the net (edge)
    - Straighter shots

**Overall:** Goals are made more often from shots made closer to the net and at straighter angles

```
Call:
qlm(formula = goal \sim x_from_edge + abs(y_fixed) + shot_angle,
   family = "binomial", data = data)
Coefficients:
             Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.719628
                        0.040596 -17.73
                                           <2e-16 ***
x_from_edge -0.033677
                        0.001092 -30.83
                                           <2e-16 ***
                        0.001976 -15.33 <2e-16 ***
abs(y_fixed) -0.030293
                        0.000822 -12.46
                                          <2e-16 ***
shot_angle
            -0.010243
               0 '*** 0.001 '** 0.01 '* 0.05 '. ' 0.1 ' 1
Signif. codes:
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 63970 on 122348
                                    degrees of freedom
Residual deviance: 59827 on 122345 degrees of freedom
  (38211 observations deleted due to missingness)
AIC: 59835
Number of Fisher Scoring iterations: 6
```

## Research Question 2

How do NHL teams differ in their <u>ability to convert shots</u> into goals, and how does this variation in goal conversion rates <u>correlate</u> with shot accuracy and effectiveness across different game periods?

# **Data Wrangling**

#### **Goal Conversion Rate**

```
team_conversion <- hockey |>
  group_by(event_team) |>
  summarise(goal_conversion = sum(shot_outcome == "GOAL") / n())
```

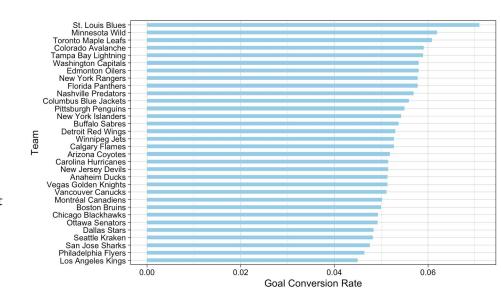
#### **Shot Accuracy Rate**

```
hockey_data <- hockey |>
  mutate(shot_accuracy = ifelse(shot_outcome == "GOAL", 1, 0)) |>
  filter(period %in% c(1, 2, 3))

accuracy_by_period <- hockey_data |>
  group_by(event_team, period) |>
  summarise(accuracy = mean(shot_accuracy))
```

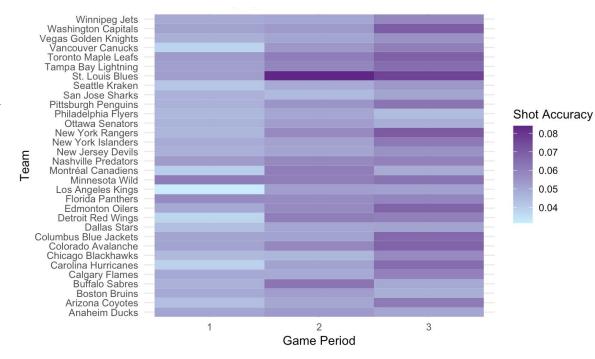
# Goal Conversion Rates by Team

- Goal conversion rates as a metric for evaluating offensive efficiency
- Insight into:
  - Shot accuracy
  - Effectiveness
  - Offensive performance
- Direct correlation between goal conversion rates and offensive performance in NHL teams.
- Teams with high conversion rates demonstrate effectiveness in turning shot opportunities into goals, suggests presence of efficient offensive strategies/skilled players.
- Lower conversion rate of teams like the Los Angeles Kings may indicate areas for improvement in their offensive gameplay.



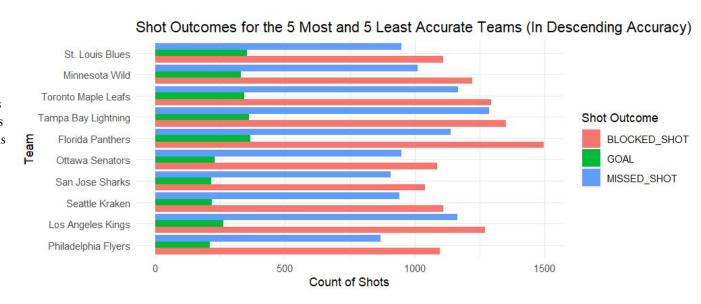
## **Shot Accuracy by Team and Period**

- Variability in teams' abilities to maintain consistent accuracy throughout a game
- Some insights:
  - St. Louis Blues exhibit exceptionally high shot accuracy in period 2
  - Los Angeles Kings demonstrate variability in shot accuracy throughout the game
  - Arizona Coyotes demonstrate progressive increase in shot accuracy
  - San Jose Sharks experience fluctuations, with period 2 showing lowest accuracy
  - Washington Capitals, Edmonton Oilers, Columbus Blue Jackets, and Colorado Avalanche display stronger shot accuracy in period 3
- Enables adaptation of strategies and tactics accordingly
- Opportunities for targeted interventions during periods of lower accuracy



## **Shot Outcomes by Team Accuracy**

- Decided to only use 10 teams in total
- Some insights:
- Teams with higher accuracies had stronger performances in terms of proportion of goals overall, as well as fewer blocked shots and missed shots
- Blues have higher proportion of goals than Flyers, and more goals as well
- Opportunities to analyze specific areas of improvement, whether it is whether your shots are blocked or missed
- Teams exhibit similar trends in shot conversion, shot accuracy, and proportion of each category of shots (Flyers have lowest accuracy + second-lowest GC)
- Combining this data with RQ1 may lead to targeted improvement for teams in several ways (location on ice, angle, player formation)



## **Research Question 3**

Do there exist any significant differences between the score distributions of home vs. away games and by what amount?

#### Motivation:

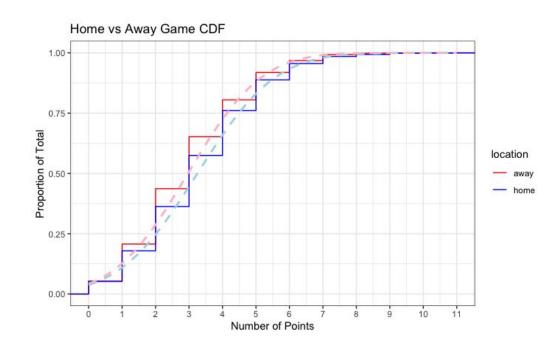
- Recently coming out of the pandemic
- Lower league wide fan attendance
- Many players temporarily prevented from playing the game

#### Home versus Away Cumulative Point Densities

#### Main insights:

- Teams score between 0 and 3 points in most games
- Both home and away score distributions followed the normal curve
- Fitted normal distribution curve for home games (in light blue) lies to the right of away games (in pink)

Seems to indicate that home games still have a higher average score than their away counterparts



## Home vs Away Distributions

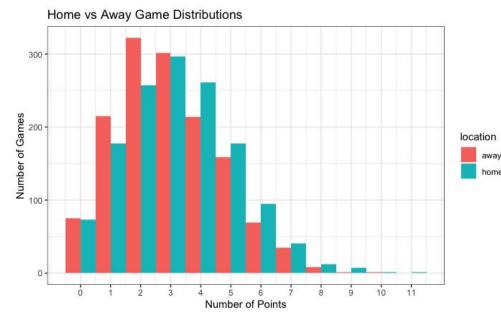
#### Main insights:

- Higher mode for home games
  - 2 points for away games
  - 3 points for home games
- Higher maximum for home games
- Smaller variance in away games than in home games

Using a t-test to examine differences in means:

- Differential means:
  - Home: 3.25 Away: 2.97
- 95% Confidence Interval of Difference in Means: 0.15 0.41
- P-value: 2.683 x10^-5

Allows us to compare games pre-pandemic to see if there are any differences in the magnitude of this separation in scores



#### Conclusions

- Discovered results that may be used to inform team managers, general managers, and offensive and defensive coordinators
- Targeted and localized results in terms of shot outcome by location on the ice, and a team's individual performance may be impacted by even where they play
- Realized trends in goal conversion and shot accuracy, identifying areas of overall performance that may be optimized
- Quantified existence of a "home-field advantage", suggesting teams should expand their outreach to more rinks

## **Future Topics**

#### Optimizing Shot Placement on the Ice

- Building on our investigation of shot location and its impact on goal-scoring likelihood
- Develop model through exploration of specific zones or angles within rink where shots are most effective
- Consider variables such as goalie positioning and defensive strategies

#### **Analyzing Player Positioning and Movement Patterns**

- Examine how player positioning and movement patterns contribute to creating goal-scoring opportunities
- Track player movements during game to understanding positioning relative to goal-scoring opportunity
- Insights into effective offensive strategies based on player positioning and movement patterns

#### **Examining Goalie Performance Metrics**

- Explore goalie performance metrics in relation to shot outcome to uncover defensive strategies
- Collect data on goalie positioning, reaction times, and block percentage in different game situations
- Conduct statistical analysis to identify correlations between goalie performance and defensive success
- Understand effective goalie strategies/positioning techniques to optimize team defensive capabilities