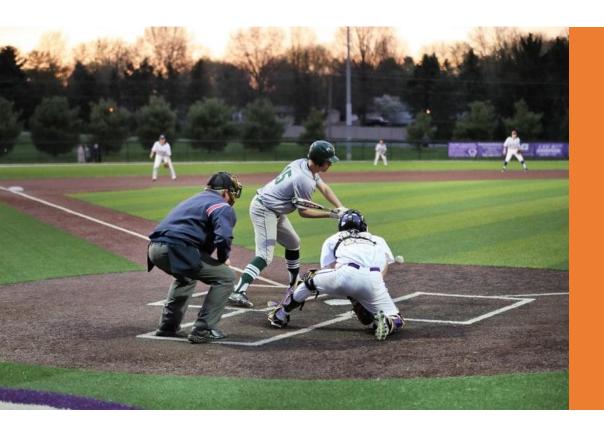


#### Introduction and Problem Statement

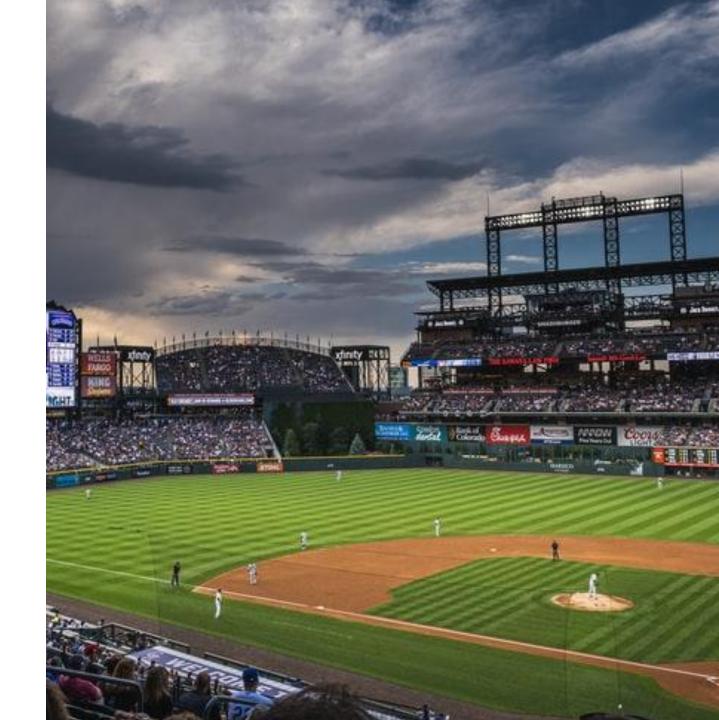


In baseball, a pitcher's success is heavily determined on the outcomes of a batted ball, and a swing and miss. Finding ways to increase success can come in many forms, velocity, spin rate, horizontal, and vertical break all play a key role.

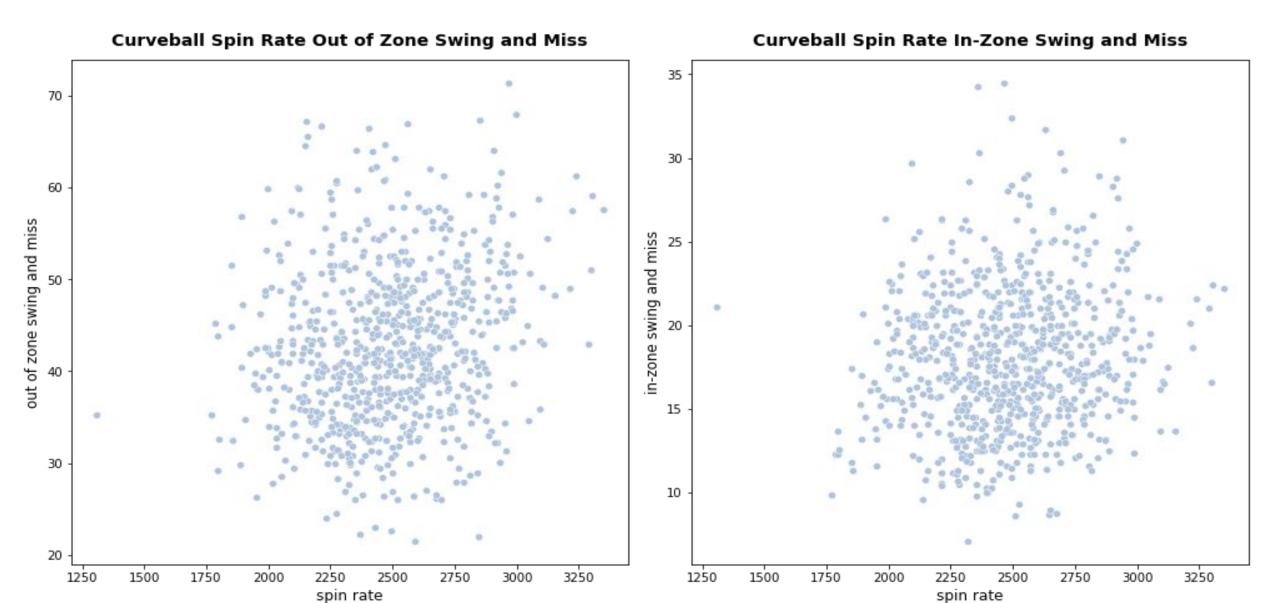
I've used advanced sabermetrics from pitcher's from the past three seasons (2019-2021) to be able to assess how to use this data to find in-game success. To do this I used an unsupervised cluster model using KMeans

#### Data Gathering and Cleaning

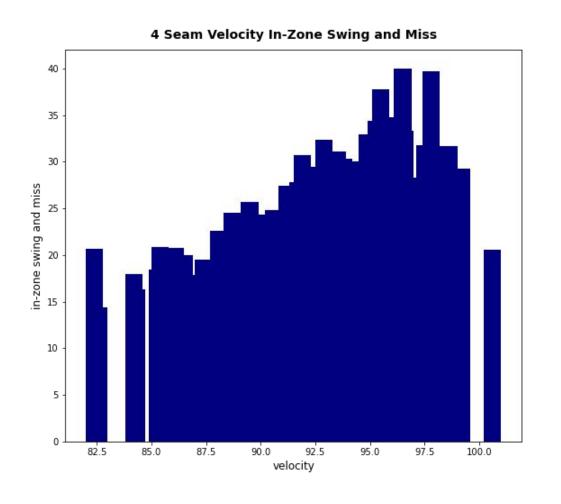
- All pitchers used had a minimum batters faced of 200 for the 2019 and 2021 data and in the 2020 data minimum of 75 batters faced due to a shortened season,
- pitch types: 4 seam fastball, slider, curveball, changeup, and cutter.
- Imputed nulls with 0

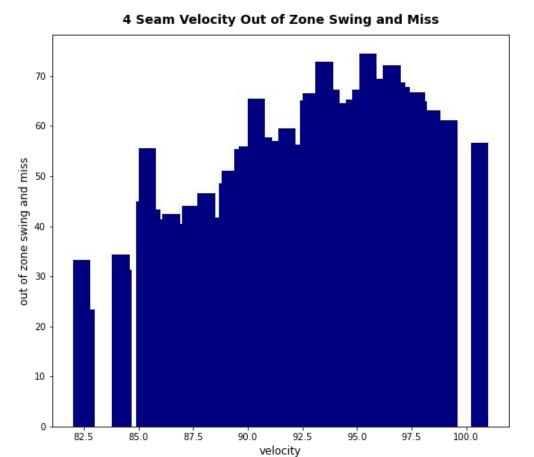


#### Fastball Velocity and Spin Rate

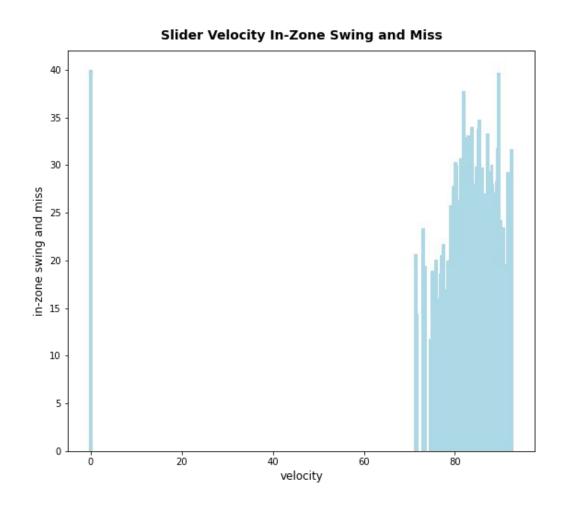


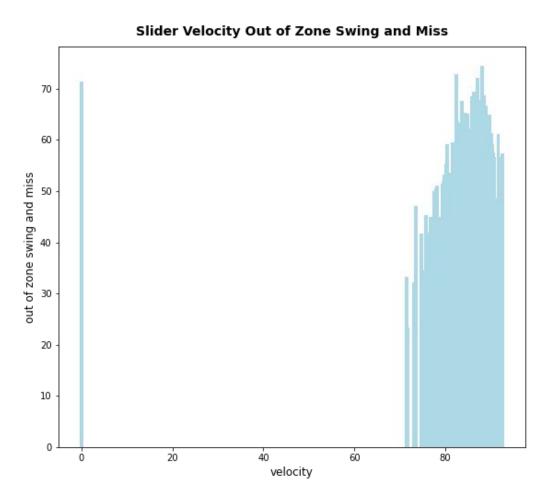
## Fastball Velocity and Spin Rate



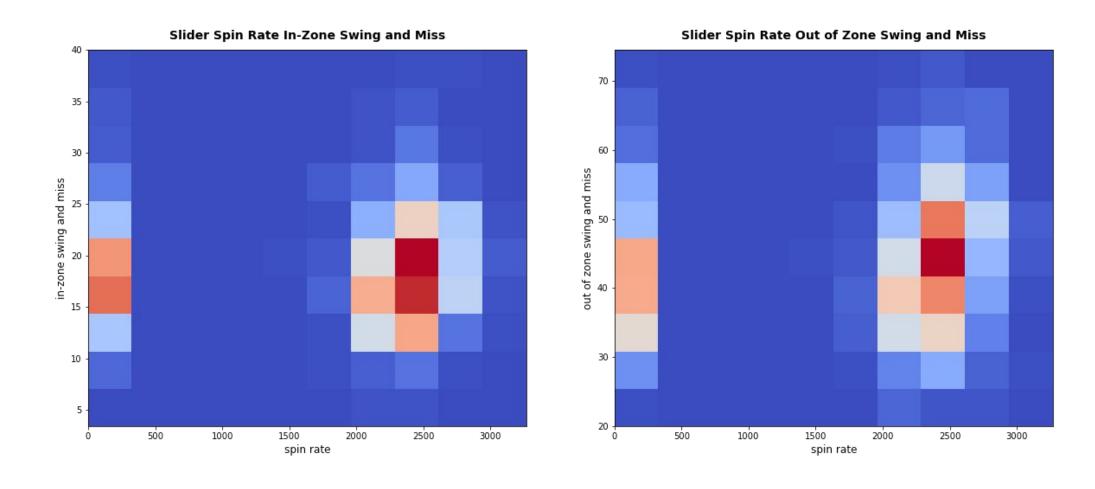


## Slider Velocity and Spin Rate

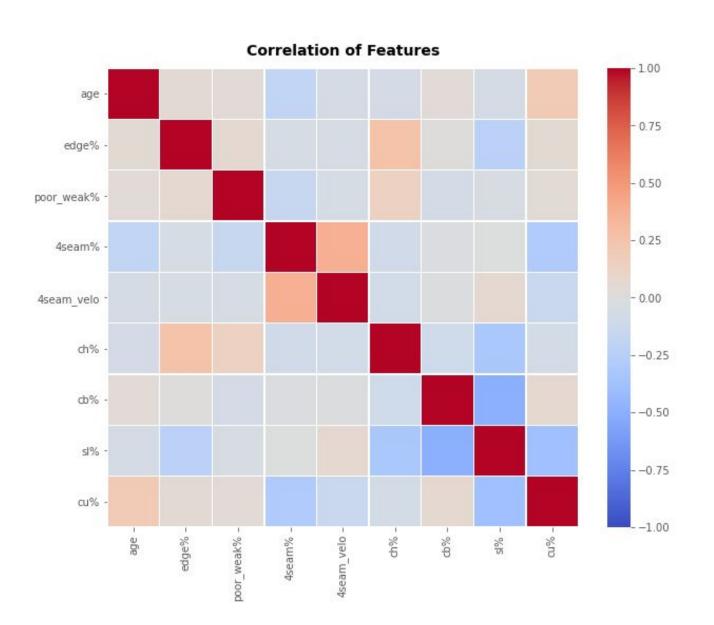




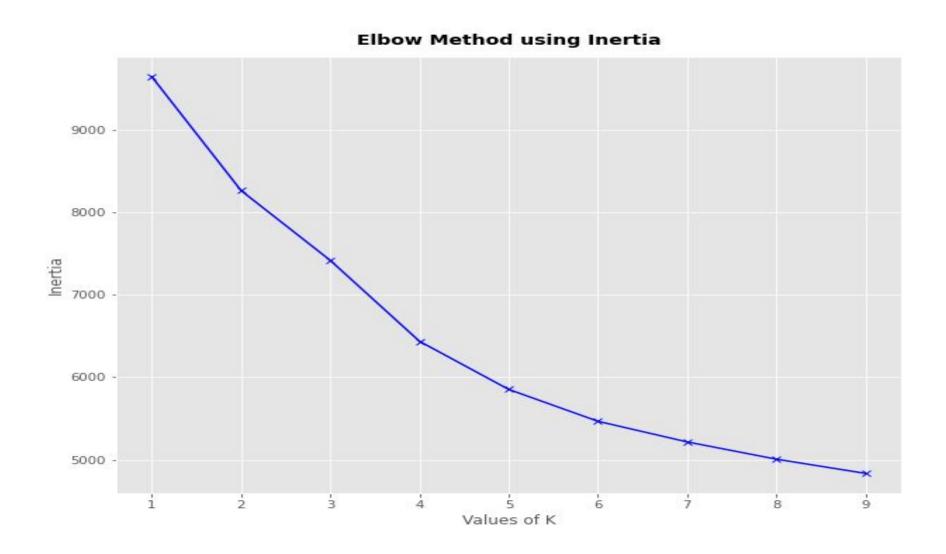
## Slider Velocity and Spin Rate



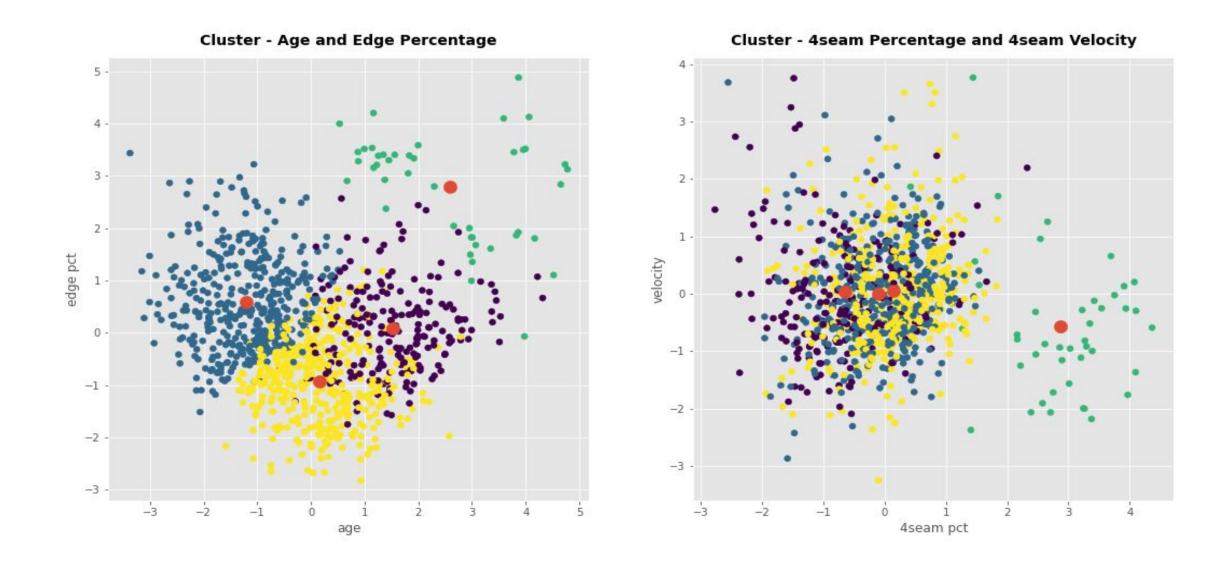
## Heatmap of Features



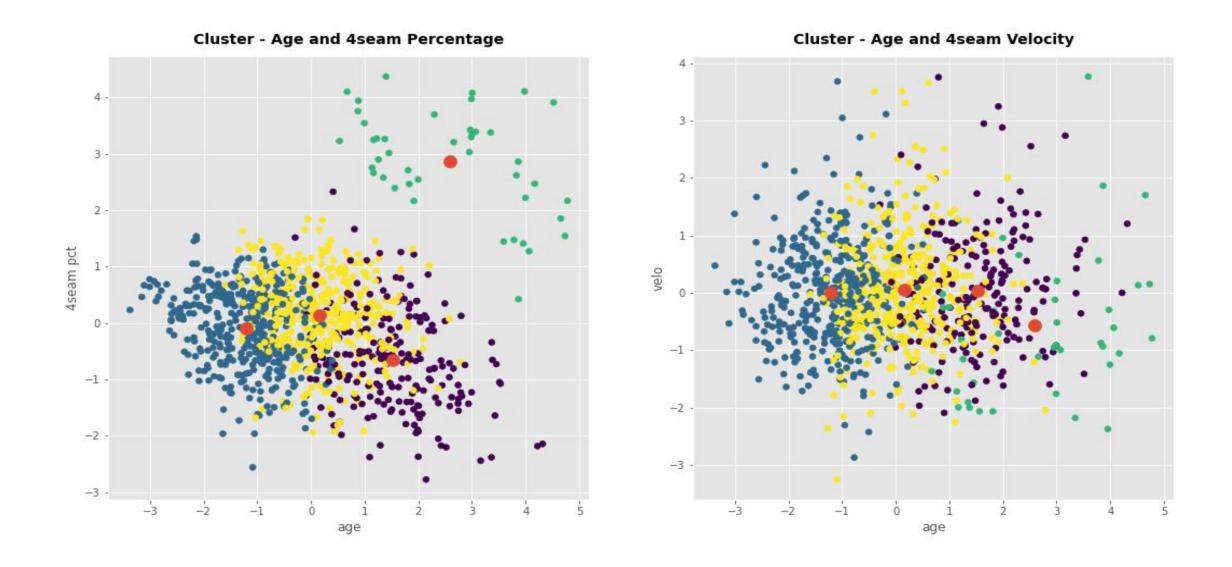
# **Determining Clusters**



#### Model Performance

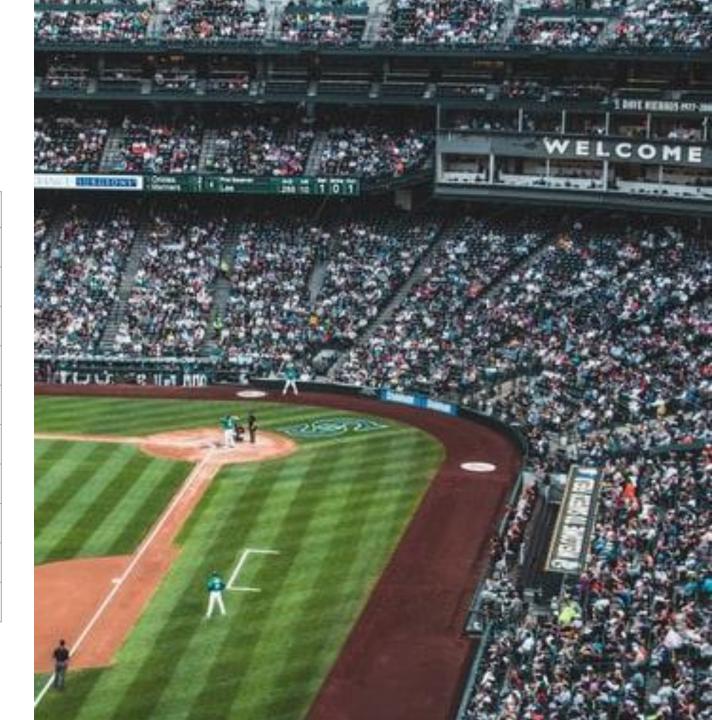


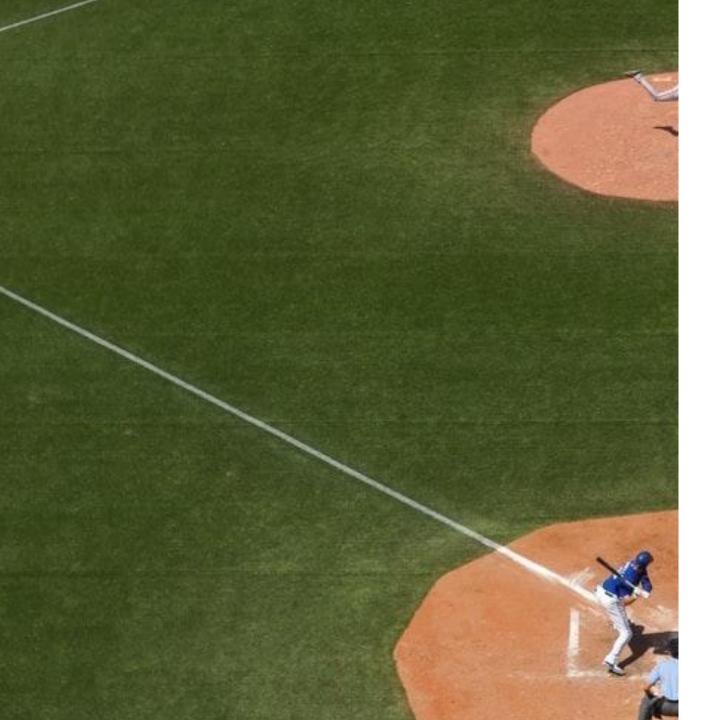
#### **Model Performance**



#### **Model Evaluation**

label	Cluster 1	Cluster 2	Cluster 3	Cluster 4
age	31.64	28.8	29.33	27.81
edge %	42.74	41.52	42.68	43.2
poor/weak %	4.42	4.07	4.67	4.27
4seam %	22.34	37	0	43.78
4seam velo	92.32	94.02	0	93.18
ch %	9.4	5.73	13.37	16.63
sl %	4.82	33.2	14.76	10.72
cu %	26.4	0.93	15.27	2.62
cb %	14.95	4.34	10.13	14.64
throws	0.68	0.79	0.53	0.7





#### **Conclusion & Recommendations**

The generalizations I made were to control what you can control. Instead of attacking hitters with their weaknesses, attack with your strengths.