

# Why is this important?

General Managers need to make many decisions:

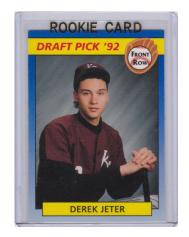
- Who are my best players now?
- What will my team look like in 5 years?
- What are my team's biggest weaknesses?
- Who should I trade from my team? & Who should I trade for?
- Who should I not trade?
  - Babe Ruth, Jeff Bagwell & Miguel Cabrera were all traded before the age of 25 and became great.

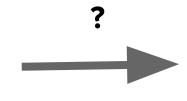
# Methodology/Workflow

- Data was scraped from baseball-reference.com
- Data includes 1151 batters (no pitchers!)
  - o Data includes all players who were playing some sort of professional baseball in 2000.
- Conclusions derived from 3 Linear regression Models: Lasso, Ridge, OLS

# Target variable: MLB At-bats Per Season

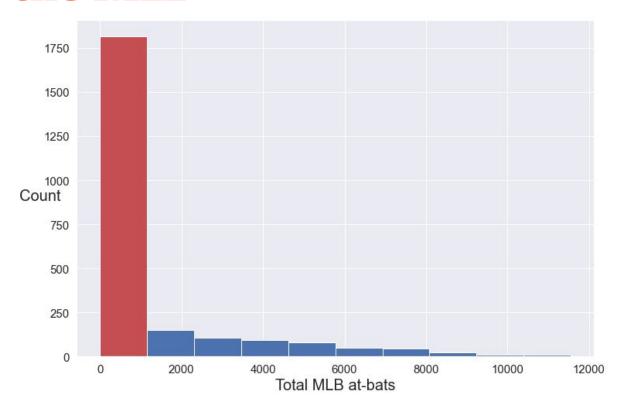
- How good a player is not easily encapsulated in a single statistic
- GMs want players that will not only be good but be consistently good for a long time.
- At-bats is a good way to measure all-around good players.







# Initial Roadbump: most players don't make the MLB



# **Findings and Conclusions**

Final R squared: 0.40

Three statistics had the highest correlation with MLB ABs:

- 1. Strikeouts per AB
- 2. OPS
- 3. Runs per AB

Feature	Impact Rank			
Debut Age	1			
Strikeouts per AB	2			
OPS	3			
Runs per AB	4			
School	5			
Draft	6			
ВМІ	7			
Position = Catcher	8			
Position = Shortstop	9			
Bats = Both	10			
Position = First Baseman	11			
Position = Second				
Baseman	12			
Bats = Left	13			
Position = Outfielder	14			
Bats = Right	15			
Position = Third				
Baseman	16			

## Findings and Conclusions Part 2

It was just as interesting to find out that the following had low impact:

- 1. Round player was drafted
- 2. Player position
- 3. Left handed or right handed

Feature	Impact Rank			
Debut Age	1			
Strikeouts per AB	2			
OPS	3			
Runs per AB	4			
School	5			
Draft	6			
ВМІ	7			
Position = Catcher	8			
Position = Shortstop	9			
Bats = Both	10			
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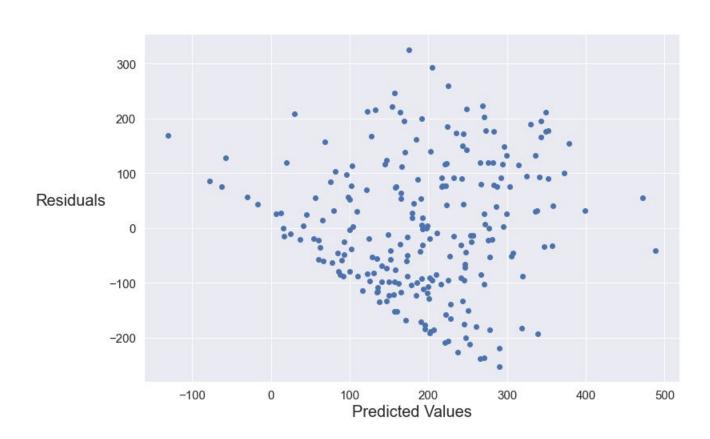
#### **Future Work**

- 1. Make a classification model for whether or not player will make it to pros
- 2. Add fielding stats.
- 3. Apply this pitchers as well



# **Appendix**

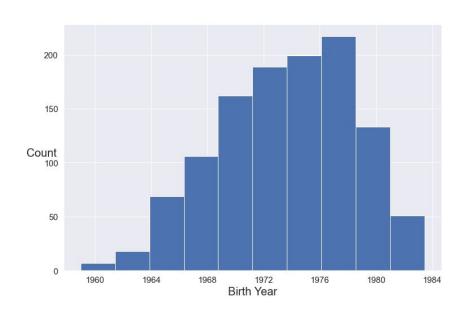
### **Residual Plot**

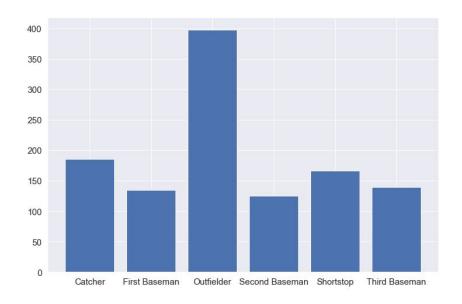


# **Summary of Regression Results**

Data with Absolute Value Coefficients											
Feature	Lasso	Ridge	OLS	Lasso Rank	Ridge Rank	OLS Rank	Average Rank	Average Coefficient	Rank of Coeff		
Strikeouts per AB	39.33956444	39.3433013	749.7253891	2	2	2	2	276.136085	1		
Runs per AB	17.32502936	17.32597847	764.6245963	4	4	1	3	266.4252014	2		
OPS	25.95188962	25.95565996	334.0712593	3	3	6	4	128.659603	3		
Bats = Both	3.596741472	2.953023042	373.929191	9	10	3	7.333333333	126.8263185	4		
Bats = Left	0.484165822	1.315538935	370.1571529	15	13	4	10.66666667	123.9856192	5		
Bats = Right	0	0.909951191	363.3325669	16	15	5	12	121.4141727	6		
Position = Shortstop	3.229632735	3.185978124	197.4597006	10	9	7	8.666666667	67.95843715	7		
Position = Second Baseman	1.806553401	1.76862869	187.6827176	12	12	8	10.66666667	63.75263323	8		
Position = First Baseman	2.506231266	2.467183741	186.2457017	11	11	9	10.33333333	63.73970556	9		
Position = Third Baseman	0.6844002821	0.7357271305	183.8767113	14	16	10	13.33333333	61.7656129	10		
Position = Outfielder	0.8527063189	0.9215998561	182.4739026	13	14	11	12.66666667	61.41606959	11		
Position = Catcher	4.80070234	4.855989016	169.680177	8	8	12	9.33333333	59.77895613	12		
Debut Age	71.48177651	71.48616173	30.26774724	1	1	14	5.333333333	57.74522849	13		
School	16.77717591	16.78983972	36.00470411	5	5	13	7.666666667	23.19057325	14		
ВМІ	10.23882379	10.24652659	5.322494446	7	7	15	9.666666667	8.602614939	15		
Draft	11.1728033	11.18358342	0.8849306858	6	6	16	9.333333333	7.747105802	16		

# **Understanding Our Sample**





#### **Actual vs. Predicted Values**

