Baez-Pitch-Selection

February 4, 2025

```
[1]: ## Chris Kellogg
   ##
   ## Python Refresher
   ##
   ##
   ## This is a refresher of data analysis and visualization using Python.
   ## With the dataset that you choose, perform the following steps:
   ## 1. Write a summary of your data and identify at least two questions to
       explore visually with your data.
   ##
   ## 2. Create a histogram or bar graph from your data.
   ##
   ## 3. Create a boxplot from your data.
   ## 4. Create a bivariate plot from your data.
   ## 5. Create any additional visualizations that will help to answer the
   ##
       question(s) you want to answer.
   ##
   ## 6. Summarize your results and make a conclusion. Explain how you arrived
       at this conclusion and how your visualizations support your conclusion.
   ##
```

```
[2]: ##
    ## load necessary packages
    ##

# import and alias Pandas
import pandas as pd
```

```
# import and alias NumPy
import numpy as np

# import and alias the plotting package
import matplotlib.pyplot as plt

# import from pybaseball API
import pybaseball
from pybaseball import playerid_lookup
from pybaseball import statcast_batter
```

```
[3]: ##
     ## read and clean up Javy Báez's Season Batting Stats
     ## -- downloaded from Baseball Reference web page
     ## -- https://www.baseball-reference.com/players/b/baezja01.shtml
     ##
     # read the Season Batting Stats file
     seasons = pd.read_csv('BaezSeasonBatting.csv')
     # drop the sub-total rows from the year he was traded
     seasons = seasons.query(
         'Year not in (2014, 2021) or Tm == "TOT"'
     )
     # keep only the fields we want
     seasons = pd.DataFrame(seasons[[
         'Year',
         'PA',
         'AB',
         'HR',
         'BB',
         'SO',
         'BA',
         'OBP',
         'SLG',
         'OPS'.
         'OPS+'
     ]])
     # preview the season batting stats
     seasons
```

```
[3]:
              PA
                   AB HR BB
                              SO
                                     BA
                                           OBP
                                                 SLG
                                                       OPS OPS+
        Year
    1
        2015
              80
                  76 1
                          4
                              24 0.289 0.325 0.408 0.733
                                                             102
    2
        2016 450
                  421 14 15 108
                                 0.273 0.314
                                               0.423 0.737
                                                              94
                  469 23 30
                              144 0.273 0.317 0.480 0.796
        2017
             508
                                                             102
```

```
4
        2018 645
                   606
                        34
                            29 167 0.290 0.326 0.554 0.881
                                                                  129
        2019
              561
                   531
                        29
                            28
                                156 0.281 0.316
                                                   0.531 0.847
    5
                                                                  115
        2020
              235
                   222
                         8
                            7
                                 75
                                     0.203 0.238
                                                   0.360 0.599
                                                                   59
              547
                                    0.265 0.319
                                                   0.494 0.813
        2021
                   502
                        31
                            28
                                184
                                                                  117
    10
        2022
              590
                   555
                        17
                            26
                                147
                                    0.238 0.278 0.393 0.671
                                                                   91
        2023
                                     0.222 0.267
                                                   0.325 0.593
    11
              547
                   510
                         9
                            24
                                125
                                                                   63
    12
        2024 289
                   272
                         6 12
                                 69 0.184 0.221 0.294 0.516
                                                                   44
[4]: ##
     ## get Javy Báez's MLBAM player ID
     ##
     # look up player id
    playerid_lookup('Báez', 'Javier')
    Gathering player lookup table. This may take a moment.
[4]:
      name_last name_first key_mlbam key_retro key_bbref key_fangraphs \
                               595879 baezj001 baezja01
           báez
                                                                   12979
                    javier
       mlb_played_first mlb_played_last
    0
                 2014.0
                                  2024.0
[5]: ##
     ## get a list all regular-season pitches thrown to Javy Báez in his career
     ##
     # get data from the pybaseball API
    javy_baez_career = statcast_batter('2015-01-01', '2024-12-31', 595879)
     # keep only the pitches thrown during regular-season games
    pitches = javy_baez_career.query('game_type == "R"')
     # keep only the fields we want
    pitches = pd.DataFrame(pitches[[
         'game_year',
         'pitch_type',
         'release_speed',
         'release_spin_rate',
         'description',
         'zone',
         'type',
         'bb_type',
         'plate_x',
         'plate_z'
    ]])
```

rename the year column

```
pitches.rename(columns={"game_year": "Year"}, inplace=True)
# mark pitches as fastballs and curveballs where appropriate
pitches['fastball'] = pitches['pitch_type'].apply(
    lambda x: 1 if x in ("FF", "SI", "FC") else 0
pitches['breaking_ball'] = pitches['pitch_type'].apply(
    lambda x: 1 if x in ("CU", "KC", "CS", "SL", "ST", "SV") else 0
# drop pitches where important info is missing
pitches.dropna(subset=[
    'pitch_type',
    'release_speed',
    'zone',
    'plate_x',
    'plate_z',
    'release_spin_rate'
], inplace=True)
# preview the pitches
pitches
```

Gathering Player Data

[5]:		Year	pitch	_type	relea	.se_speed	release	e_spin_rat	e d	lescription	\
	38	2020		CH		85.9		1534.) hit	_into_play	
	39	2020		SI		89.5		2153.)	ball	
	40	2020		CH		85.5		1629.) cal	led_strike	
	41	2020		FF		95.0		2202.)	ball	
	42	2020		FF		94.9		2127.)	foul	
	•••	•••	•••		•••		•	••			
	7591	2021		SI		92.1		1854.) swing	ging_strike	
	7592	2021		FF		93.1		2131.)	ball	
	7593	2021		FF		96.9		2231.)	ball	
	7594	2021		CU		74.0		2314.) swing	ging_strike	
	7595	2021		CU		77.5		2241.) cal	.led_strike	
		zone	type	bb	_type	plate_x	plate_z	z fastbal	l break	ing_ball	
	38	12.0	Х	line_	drive	1.05	2.81	1)	0	
	39	12.0	В		NaN	1.25	3.38	3	1	0	
	40	9.0	S		NaN	0.62	1.88	3 ()	0	
	41	14.0	В		NaN	1.63	1.77	7	1	0	
	42	7.0	S		NaN	-0.46	2.11	1	1	0	
	•••			•••	•••	•••	•••				
	7591	4.0	S		NaN	-0.34	2.40		1	0	
	7592	11.0	В		NaN	-0.20	4.94	1	1	0	

```
7593 12.0
                                               3.05
                                                                               0
               В
                           {\tt NaN}
                                    1.16
                                                              1
7594 14.0
                S
                           {\tt NaN}
                                     1.31
                                               1.30
                                                              0
                                                                               1
7595 1.0
                S
                                   -0.47
                                               2.89
                                                              0
                           {\tt NaN}
                                                                               1
```

[15715 rows x 12 columns]

```
[6]:
       Year pitch_count
    0 2015
                    279
    1 2016
                    1510
    2 2017
                   1784
    3 2018
                   2160
    4 2019
                   1935
    5 2020
                    905
    6 2021
                   2069
    7 2022
                   2092
    8 2023
                    1961
    9 2024
                    1020
```

```
fastball_data
[7]:
       Year mean_fb_velo fb_count
    0 2015
                93.405263
                                171
    1 2016
                92.846911
                                874
    2 2017
                93.121673
                                992
    3 2018
                92.904718
                               1187
    4 2019
                93.051683
                               1099
    5 2020
                92.624580
                                476
    6 2021
                93.222592
                               1111
    7 2022
                92.772956
                                954
    8 2023
                93.215733
                               1017
    9 2024
                93.280591
                                474
[8]: ##
     ## get the aggregate info for breaking balls
     # get the mean spin rate and count for each year
    breaking_ball_data = pitches \
         .query('breaking_ball == 1') \
        .groupby(['Year']) \
         .agg(
            mean_bb_spin = ('release_spin_rate', 'mean'),
            bb_count = ('pitch_type', 'count')
        ) \
         .reset_index()
     # preview the breaking ball data
    breaking_ball_data
[8]:
       Year mean_bb_spin bb_count
    0 2015
              2141.298701
                                 77
    1 2016
              2343.246696
                                454
    2 2017
              2378.603741
                                588
    3 2018
              2404.890039
                                773
    4 2019
              2485.948678
                                643
    5 2020 2545.185915
                                355
    6 2021
              2473.761836
                                697
    7 2022
              2467.678490
                                874
    8 2023
              2455.187679
                                698
    9 2024
              2487.671946
                                442
[9]: ##
     ## merge the aggregates to the season data
     ##
```

preview the fastball data

```
# add pitch data onto the season data
df = pd.merge(
    seasons,
    pitch_data,
    how = 'left',
    on = ['Year']
)
# add fastball data onto the season data
df = pd.merge(
    df,
    fastball_data,
    how = 'left',
    on = ['Year']
)
# add breaking ball data onto the list of 1000-pitch seasons
df = pd.merge(
    df,
    breaking_ball_data,
    how = 'left',
    on = ['Year']
)
# add columns for fastballs and breaking balls as percentage
df['fb_percent'] = df.fb_count / df.pitch_count
df['bb_percent'] = df.bb_count / df.pitch_count
# keep only the fields we want
df = pd.DataFrame(df[[
    'Year',
    'PA',
    'BA',
    'OBP',
    'OPS',
    'OPS+',
    'mean_fb_velo',
    'mean_bb_spin',
    'pitch_count',
    'fb_count',
    'fb_percent',
    'bb_count',
    'bb_percent'
]])
df
```

```
[9]:
         Year
                PA
                       BA
                              OBP
                                     OPS
                                          OPS+
                                                mean_fb_velo
                                                               mean_bb_spin \
                                                   93.405263
         2015
                   0.289
                           0.325
                                           102
                                                                2141.298701
      0
                80
                                   0.733
      1 2016 450
                   0.273
                           0.314
                                   0.737
                                            94
                                                   92.846911
                                                                2343.246696
         2017
               508
                    0.273
                            0.317
                                   0.796
                                           102
                                                   93.121673
                                                                2378.603741
         2018
      3
               645
                    0.290
                            0.326
                                   0.881
                                           129
                                                   92.904718
                                                                2404.890039
      4
         2019
               561
                   0.281
                            0.316
                                   0.847
                                           115
                                                   93.051683
                                                                2485.948678
         2020
               235
                    0.203
                            0.238
                                   0.599
                                            59
                                                   92.624580
                                                                2545.185915
       2021
      6
               547 0.265
                           0.319
                                   0.813
                                           117
                                                   93.222592
                                                                2473.761836
      7
         2022 590 0.238
                           0.278
                                   0.671
                                                   92.772956
                                                                2467.678490
                                            91
      8 2023
               547
                    0.222
                            0.267
                                   0.593
                                            63
                                                   93.215733
                                                                2455.187679
         2024
      9
               289
                   0.184
                           0.221
                                   0.516
                                            44
                                                   93.280591
                                                                2487.671946
         pitch_count
                     fb_count
                               fb_percent
                                             bb_count
                                                       bb_percent
      0
                 279
                                   0.612903
                                                   77
                                                          0.275986
                            171
                1510
                            874
                                                  454
      1
                                   0.578808
                                                          0.300662
      2
                1784
                            992
                                   0.556054
                                                  588
                                                          0.329596
      3
                2160
                          1187
                                   0.549537
                                                  773
                                                          0.357870
      4
                1935
                           1099
                                   0.567959
                                                  643
                                                          0.332300
      5
                 905
                            476
                                   0.525967
                                                  355
                                                          0.392265
      6
                2069
                          1111
                                   0.536974
                                                  697
                                                          0.336878
      7
                2092
                            954
                                   0.456023
                                                  874
                                                          0.417782
      8
                1961
                           1017
                                                   698
                                                          0.355941
                                   0.518613
      9
                1020
                            474
                                   0.464706
                                                   442
                                                          0.433333
[10]: ##
      ## subset the data for just the specific question we want to address
      ##
      # keep only the fields we want
      df = pd.DataFrame(df[[
          'Year',
          'OPS+',
          'bb_percent'
      ]])
      df
[10]:
         Year
               OPS+
                     bb_percent
      0
         2015
                102
                       0.275986
         2016
      1
                 94
                       0.300662
      2 2017
                102
                       0.329596
      3 2018
                129
                       0.357870
      4 2019
                115
                       0.332300
      5 2020
                 59
                       0.392265
      6 2021
                117
                       0.336878
      7
         2022
                 91
                       0.417782
         2023
                 63
                       0.355941
```

9 2024 44 0.433333

```
Γ11]: ##
     ## 1. Write a summary of your data and identify at least two questions to
          explore visually with your data.
     ##
     # I've been a Cubs fan for about 40 years or so, and they've only won the
     # World Series once in that time. One of the most exciting players on that
     # winning team was Javier Báez. He was a good hitter, and many Cubs fans were
     # disappointed when Báez was traded by the Cubs instead of being signed to a
     # contract extension in 2021. In the three seasons since the trade, Báez has
     # been a much worse hitter than he was during his time in Chicago, often
     # looking overmatched and swinging at pitches that are unhittable. I built a
     # data set (from Baseball Reference and Statcast data) to see whether the
     # decline in performance can be explained by the selection and quality of
     # pitches thrown to Báez.
     # Q. In three years since 2021, are MLB pitchers throwing Javier Báez more
        breaking pitches, explaining his decline in performance?
```

```
##
## 2. Create a histogram or bar graph from your data.
##

# plot the OPS+ for each year
plt.bar(df['Year'], df['OPS+'])

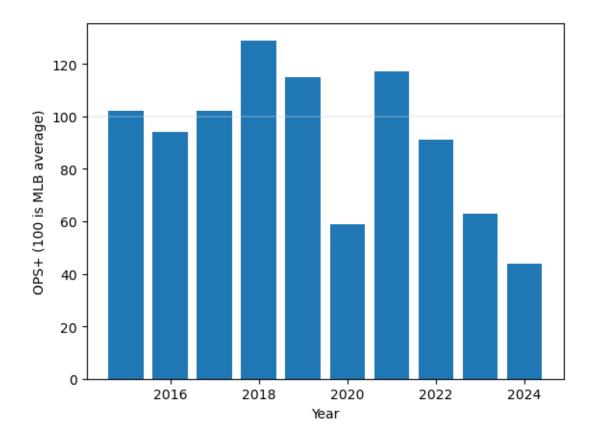
# add a chart title
plt.suptitle('OPS+ for Javy Báez over career (by year)')

# add the axis labels
plt.xlabel('Year')
plt.ylabel('OPS+ (100 is MLB average)')

# add the MLB average
plt.axhline(y = 100, color = 'lightgray', alpha=0.3)

# show the plot
plt.show()
```

OPS+ for Javy Báez over career (by year)



```
##
## 3. Create a boxplot from your data.
##

# box plot the mean fastball velocity
fig = plt.figure(figsize =(10, 7))
ax = fig.add_subplot(111)
plt.boxplot(df.bb_percent, vert = 0)

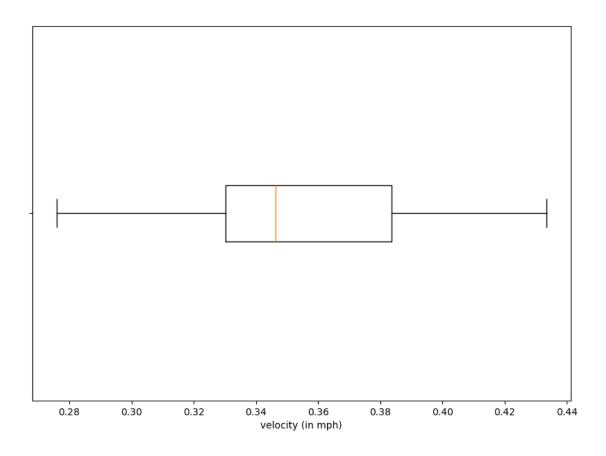
# add a chart title
plt.suptitle('Breaking pitch percentage for Javy Báez (year over year)')

# remove y-axis labels
ax.set_yticklabels([''])
plt.ylabel('')

# add the x-axis label
plt.xlabel('velocity (in mph)')
```

```
# show plot
plt.show()
```

Breaking pitch percentage for Javy Báez (year over year)



```
##
## 4. Create a bivariate plot from your data.
##

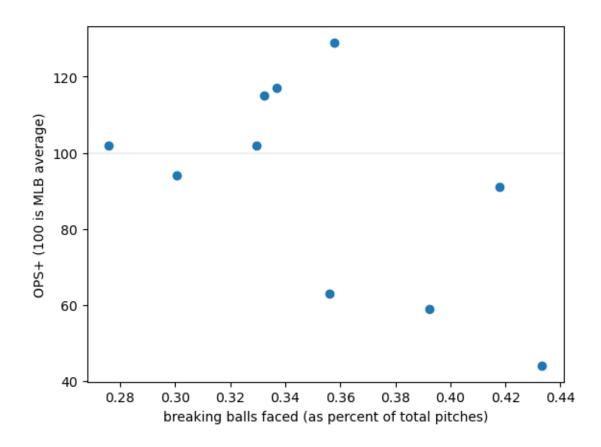
# plot the percentage of breaking balls against OPS+
plt.scatter(df['bb_percent'], df['OPS+'])

# add a chart title
plt.suptitle('Percentage of breaking balls vs. Baez OPS+')

# add the x-axis label
plt.xlabel('breaking balls faced (as percent of total pitches)')
plt.ylabel('OPS+ (100 is MLB average)')
```

```
# add the MLB average
plt.axhline(y = 100, color = 'lightgray', alpha=0.3)
# show plot
plt.show()
```

Percentage of breaking balls vs. Baez OPS+



```
[15]: ##
    ## 5. Create any additional visualizations that will help to answer the
    ## question(s) you want to answer.
##

# plot the percentage of breaking balls in each year
plt.scatter(df['Year'], df['bb_percent'])

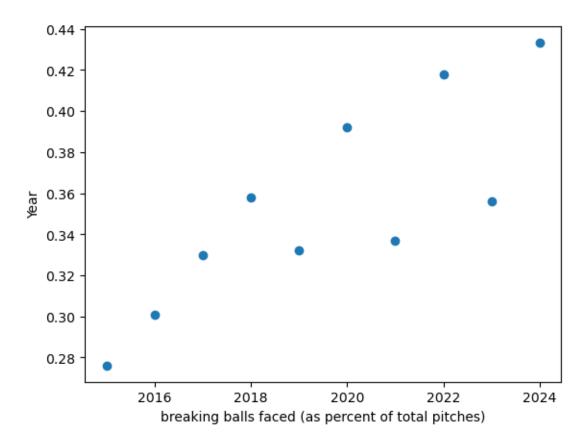
# add a chart title
plt.suptitle('Percentage of breaking balls in each year')

# add the x-axis label
```

```
plt.xlabel('breaking balls faced (as percent of total pitches)')
plt.ylabel('Year')

# show plot
plt.show()
```

Percentage of breaking balls in each year



```
## ## 6. Summarize your results and make a conclusion. Explain how you arrived
## at this conclusion and how your visualizations support your conclusion.
##

# Based on the bar chart above, we can see that Báez's OPS+ (sort of an
# overall measure of offensive value, normalized against the mean performance
# of all MLB hitters) has been declining sharply since the 2021 season, when
# he was traded from the Cubs. From the box-and-whisker plot, we can see that
# the percentage of breaking balls thrown by MLB pitcher to Báez has varied
# widely over his career, and from the first scatter plot, it's pretty evident
# that Báez generally performs worse offensively when the pitch mix includes
```