

1. The Statcast revolution



This is Aaron Judge. Judge is one of the physically largest players in Major League Baseball standing 6 feet 7 inches (2.01 m) tall and weighing 282 pounds (128 kg). He also hit the [hardest home run](#) ever recorded. How do we know this? **Statcast**.

Statcast is a state-of-the-art tracking system that uses high-resolution cameras and radar equipment to measure the precise location and movement of baseballs and baseball players. Introduced in 2015 to all 30 major league ballparks, Statcast data is revolutionizing the game. Teams are engaging in an "arms race" of data analysis, hiring analysts left and right in an attempt to gain an edge over their competition. This [video](#) describing the system is incredible.

In this notebook, we're going to wrangle, analyze, and visualize Statcast data to compare Mr. Judge and another (extremely large) teammate of his. Let's start by loading the data into our Notebook.

There are two CSV files, `judge.csv` and `stanton.csv`, both of which

contain Statcast data for 2015-2017. We'll use pandas DataFrames to store this data. Let's also load our data visualization libraries, matplotlib and seaborn.

```
In [2]: !pip install seaborn
```

Collecting seaborn

Downloading seaborn-0.13.2-py3-none-any.whl.metadata (5.4 kB)

Requirement already satisfied: numpy!=1.24.0,>=1.20 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefounda

Requirement already satisfied: pandas>=1.2 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefounda.pyth

Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefou

Requirement already satisfied: contourpy>=1.0.1 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio

Requirement already satisfied: cycler>=0.10 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio.pyt

Requirement already satisfied: fonttools>=4.22.0 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio

Requirement already satisfied: kiwisolver>=1.3.1 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio

Requirement already satisfied: packaging>=20.0 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio.

Requirement already satisfied: pillow>=8 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio.pythor

Requirement already satisfied: pyparsing>=2.3.1 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio

Requirement already satisfied: python-dateutil>=2.7 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefounda

Requirement already satisfied: pytz>=2020.1 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio.pyt

Requirement already satisfied: tzdata>=2022.7 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio.p

Requirement already satisfied: six>=1.5 in c:\users\lenovo\appdata\local\packages\pythonsoftwarefoundatio.python.

Downloading seaborn-0.13.2-py3-none-any.whl (294 kB)

----- 0.0/294.9 kB ? eta -:-:-

----- 0.0/294.9 kB ? eta -:-:-

----- 10.2/294.9 kB ? eta -:-:-

----- 30.7/294.9 kB 435.7 kB/s eta 0:00:01

----- 92.2/294.9 kB 751.6 kB/s eta 0:00:01

----- 294.9/294.9 kB 1.8 MB/s eta 0:00:00

Installing collected packages: seaborn

Successfully installed seaborn-0.13.2

[notice] A new release of pip is available: 24.0 -> 24.1

[notice] To update, run:

C:\Users\LENOVO\AppData\Local\Microsoft\WindowsApps\PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0\python.exe

-m pip install --upgrade pip

```
In [3]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline

# Load Aaron Judge's Statcast data
judge = pd.read_csv('datasets/judge.csv')

# Load Giancarlo Stanton's Statcast data
stanton = pd.read_csv('datasets/stanton.csv')
```

2. What can Statcast measure?

The better question might be, what can't Statcast measure?

Starting with the pitcher, Statcast can measure simple data points such as velocity. At the same time, Statcast digs a whole lot deeper, also measuring the release point and spin rate of every pitch.

Moving on to hitters, Statcast is capable of measuring the exit velocity, launch angle and vector of the ball as it comes off the bat. From there, Statcast can also track the hang time and projected distance that a ball travels.

Let's inspect the last five rows of the judge DataFrame. You'll see that each row represents one pitch thrown to a batter. You'll also see that some columns have esoteric names. If these don't make sense now, don't worry. The relevant ones will be explained as necessary.

```
In [4]: # Display all columns (pandas will collapse some columns if we don't set this option)
pd.set_option('display.max_columns', None)

# Display the last five rows of the Aaron Judge file

print(judge.tail())
```

	pitch_type	game_date	release_speed	release_pos_x	release_pos_z	\
3431	CH	2016-08-13	85.6	-1.9659	5.9113	
3432	CH	2016-08-13	87.6	-1.9318	5.9349	
3433	CH	2016-08-13	87.2	-2.0285	5.8656	
3434	CU	2016-08-13	79.7	-1.7108	6.1926	
3435	FF	2016-08-13	93.2	-1.8476	6.0063	

	player_name	batter	pitcher	events	description	spin_dir	\
3431	Aaron Judge	592450	542882	NaN	ball	NaN	
3432	Aaron Judge	592450	542882	home_run	hit_into_play_score	NaN	
3433	Aaron Judge	592450	542882	NaN	ball	NaN	
3434	Aaron Judge	592450	542882	NaN	foul	NaN	
3435	Aaron Judge	592450	542882	NaN	called_strike	NaN	

	spin_rate_deprecated	break_angle_deprecated	break_length_deprecated	\
3431	NaN	NaN	NaN	
3432	NaN	NaN	NaN	
3433	NaN	NaN	NaN	
3434	NaN	NaN	NaN	
3435	NaN	NaN	NaN	

	zone	des	game_type	stand	\
3431	14.0	NaN	R	R	
3432	4.0	Aaron Judge homers (1) on a fly ball to center...	R	R	
3433	14.0	NaN	R	R	
3434	4.0	NaN	R	R	
3435	8.0	NaN	R	R	

	p_throws	home_team	away_team	type	hit_location	bb_type	balls	\
3431	R	NY Yankees	TB	B	NaN	NaN	0	
3432	R	NY Yankees	TB	X	NaN	fly_ball	1	
3433	R	NY Yankees	TB	B	NaN	NaN	0	
3434	R	NY Yankees	TB	S	NaN	NaN	0	
3435	R	NY Yankees	TB	S	NaN	NaN	0	

	strikes	game_year	pfx_x	pfx_z	plate_x	plate_z	on_3b	on_2b	\
3431	0	2016	-0.379108	0.370567	0.739	1.442	NaN	NaN	
3432	2	2016	-0.295608	0.320400	-0.419	3.273	NaN	NaN	
3433	2	2016	-0.668575	0.198567	0.561	0.960	NaN	NaN	
3434	1	2016	0.397442	-0.614133	-0.803	2.742	NaN	NaN	
3435	0	2016	-0.823050	1.623300	-0.273	2.471	NaN	NaN	

	on_1b	outs_when_up	inning	inning_topbot	hc_x	hc_y	\
3431	NaN	0	5	Bot	NaN	NaN	
3432	NaN	2	2	Bot	130.45	14.58	
3433	NaN	2	2	Bot	NaN	NaN	
3434	NaN	2	2	Bot	NaN	NaN	
3435	NaN	2	2	Bot	NaN	NaN	

	tfs_deprecated	tfs_zulu_deprecated	pos2_person_id	umpire	\
3431	NaN	NaN	571912.0	NaN	
3432	NaN	NaN	571912.0	NaN	
3433	NaN	NaN	571912.0	NaN	
3434	NaN	NaN	571912.0	NaN	
3435	NaN	NaN	571912.0	NaN	

	sv_id	vx0	vy0	vz0	ax	ay	az	sz_top	\
3431	160813_144259	6.960	-124.371	-4.756	-2.821	23.634	-30.220	3.93	
3432	160813_135833	4.287	-127.452	-0.882	-1.972	24.694	-30.705	4.01	
3433	160813_135815	7.491	-126.665	-5.862	-6.393	21.952	-32.121	4.01	
3434	160813_135752	1.254	-116.062	0.439	5.184	21.328	-39.866	4.01	
3435	160813_135736	5.994	-135.497	-6.736	-9.360	26.782	-13.446	4.01	

	sz_bot	hit_distance_sc	launch_speed	launch_angle	effective_speed	\
3431	1.82	NaN	NaN	NaN	84.459	
3432	1.82	446.0	108.8	27.410	86.412	
3433	1.82	NaN	NaN	NaN	86.368	
3434	1.82	9.0	55.8	-24.973	77.723	
3435	1.82	NaN	NaN	NaN	92.696	

	release_spin_rate	release_extension	game_pk	pos1_person_id	\
3431	1552.0	5.683	448611	542882.0	
3432	1947.0	5.691	448611	542882.0	
3433	1761.0	5.721	448611	542882.0	

3434	2640.0	5.022	448611	542882.0
3435	2271.0	6.068	448611	542882.0

	pos2_person_id	pos3_person_id	pos4_person_id	pos5_person_id	\
3431	571912.0	543543.0	523253.0	446334.0	
3432	571912.0	543543.0	523253.0	446334.0	
3433	571912.0	543543.0	523253.0	446334.0	
3434	571912.0	543543.0	523253.0	446334.0	
3435	571912.0	543543.0	523253.0	446334.0	

	pos6_person_id	pos7_person_id	pos8_person_id	pos9_person_id	\
3431	622110.0	545338.0	595281.0	543484.0	
3432	622110.0	545338.0	595281.0	543484.0	
3433	622110.0	545338.0	595281.0	543484.0	
3434	622110.0	545338.0	595281.0	543484.0	
3435	622110.0	545338.0	595281.0	543484.0	

	release_pos_y	estimated_ba_using_speedangle	\
3431	54.8144	0.00	
3432	54.8064	0.98	
3433	54.7770	0.00	
3434	55.4756	0.00	
3435	54.4299	0.00	

	estimated_woba_using_speedangle	woba_value	woba_denom	babip_value	\
3431	0.000	NaN	NaN	NaN	
3432	1.937	2.0	1.0	0.0	
3433	0.000	NaN	NaN	NaN	
3434	0.000	NaN	NaN	NaN	
3435	0.000	NaN	NaN	NaN	

	iso_value	launch_speed_angle	at_bat_number	pitch_number
3431	NaN	NaN	36	1
3432	3.0	6.0	14	4
3433	NaN	NaN	14	3
3434	NaN	1.0	14	2
3435	NaN	NaN	14	1

3. Aaron Judge and Giancarlo Stanton, prolific sluggers



This is Giancarlo Stanton. He is also a very large human being, standing 6 feet 6 inches tall and weighing 245 pounds. Despite not wearing the same jersey as Judge in the pictures provided, in 2018 they will be teammates on the New York Yankees. They are similar in a lot of ways, one being that they hit a lot of home runs. Stanton and Judge led baseball in home runs in 2017, with [59](#) and [52](#), respectively. These are exceptional totals - the player in third "only" had 45 home runs.

Stanton and Judge are also different in many ways. One is [batted ball events](#), which is any batted ball that produces a result. This includes outs, hits, and errors. Next, you'll find the counts of batted ball events for each player in 2017. The frequencies of other events are quite different.

In [5]:

```
# All of Aaron Judge's batted ball events in 2017
judge_events_2017 = judge.loc[judge['game_year'] == 2017]
print("Aaron Judge batted ball event totals, 2017:")
print(judge_events_2017.value_counts())

# All of Giancarlo Stanton's batted ball events in 2017
stanton_events_2017 = stanton.loc[stanton['game_year'] == 2017]
print("\nGiancarlo Stanton batted ball event totals, 2017:")
print(stanton_events_2017.value_counts())
```

Aaron Judge batted ball event totals, 2017:

```
events
strikeout                207
field_out                146
walk                    116
single                   75
home_run                 52
double                   24
grounded_into_double_play 15
intent_walk              11
force_out                11
hit_by_pitch              5
sac_fly                   4
fielders_choice_out       4
field_error               4
triple                    3
strikeout_double_play     1
Name: count, dtype: int64
```

Giancarlo Stanton batted ball event totals, 2017:

```
events
field_out      239
strikeout      161
single         77
walk           72
home_run       59
double         32
intent_walk    13
grounded_into_double_play 13
force_out      7
hit_by_pitch   7
field_error    5
sac_fly        3
fielders_choice_out 2
strikeout_double_play 2
pickoff_1b     1
Name: count, dtype: int64
```

4. Analyzing home runs with Statcast data

So Judge walks and strikes out more than Stanton. Stanton flies out more than Judge. But let's get into their hitting profiles in more detail. Two of the most groundbreaking Statcast metrics are launch angle and exit velocity:

- [Launch angle](#): the vertical angle at which the ball leaves a player's bat
- [Exit velocity](#): the speed of the baseball as it comes off the bat

This new data has changed the way teams value both hitters and pitchers. Why? As per the [Washington Post](#):

Balls hit with a high launch angle are more likely to result in a hit. Hit fast enough and at the right angle, they become home runs.

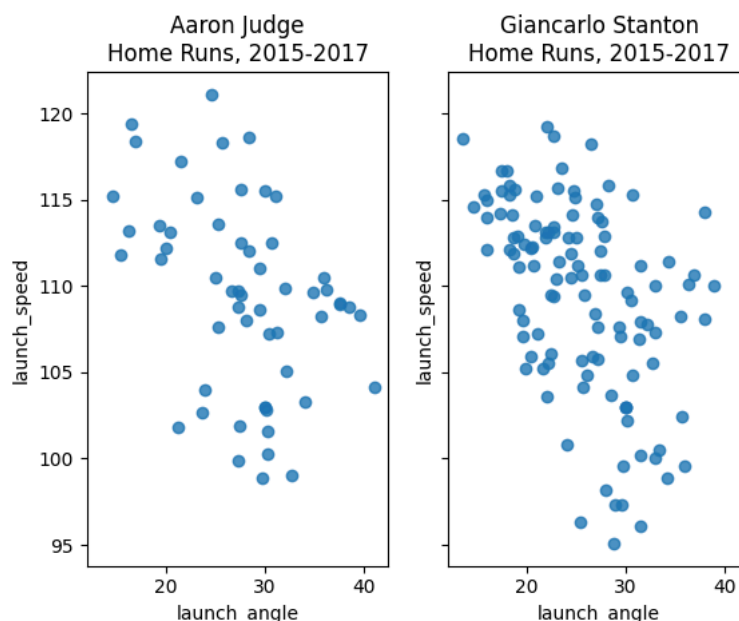
Let's look at exit velocity vs. launch angle and let's focus on home runs only (2015-2017). The first two plots show data points. The second two show smoothed contours to represent density.

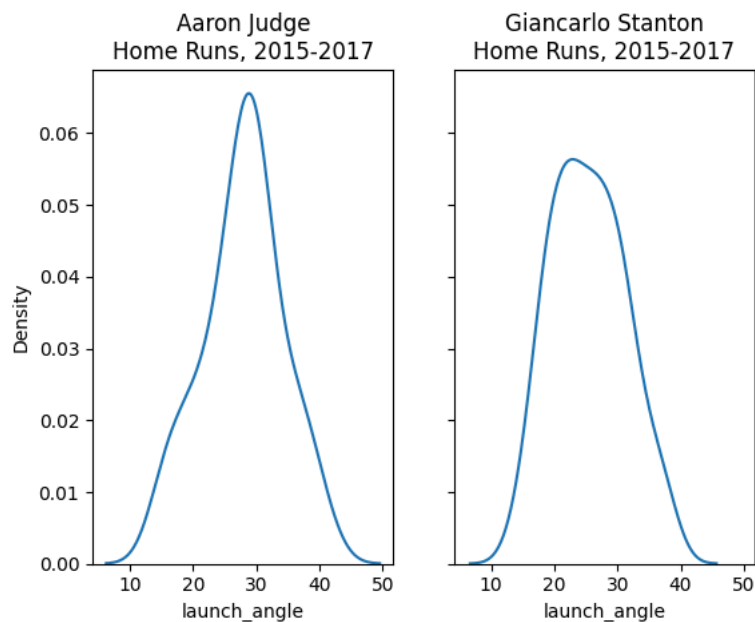
```
In [22]: # Filter to include home runs only
judge_hr = judge[judge['events']=='home_run']
stanton_hr = stanton[stanton['events']=='home_run']

# Create a figure with two scatter plots of launch speed vs. launch angle, one for ea
fig1, axs1 = plt.subplots(ncols=2, sharex=True, sharey=True)
sns.regplot(x='launch_angle', y='launch_speed', fit_reg=False, color='tab:blue', data
sns.regplot(x='launch_angle', y='launch_speed', fit_reg=False, color='tab:blue', data

# Create a figure with two KDE plots of launch speed vs. launch angle, one for each p
fig2, axs2 = plt.subplots(ncols=2, sharex=True, sharey=True)
sns.kdeplot(judge_hr.launch_angle, ax=axs2[0]).set_title('Aaron Judge\nHome Runs, 201
sns.kdeplot(stanton_hr.launch_angle, ax=axs2[1]).set_title('Giancarlo Stanton\nHome Ru
```

```
Out [22]: Text(0.5, 1.0, 'Giancarlo Stanton\nHome Runs, 2015-2017')
```





5. Home runs by pitch velocity

It appears that Stanton hits his home runs slightly lower and slightly harder than Judge, though this needs to be taken with a grain of salt given the small sample size of home runs.

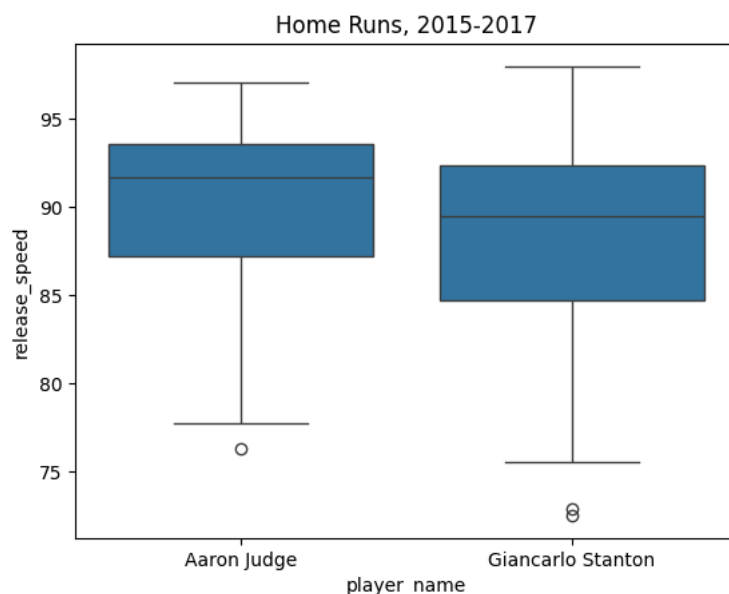
Not only does Statcast measure the velocity of the ball coming off of the bat, it measures the velocity of the ball coming out of the pitcher's hand and begins its journey towards the plate. We can use this data to compare Stanton and Judge's home runs in terms of pitch velocity. Next you'll find box plots displaying the five-number summaries for each player: minimum, first quartile, median, third quartile, and maximum.

```
In [23]: # Combine the Judge and Stanton home run DataFrames for easy boxplot plotting
judge_stanton_hr = pd.concat([judge_hr, stanton_hr])

# Create a boxplot that describes the pitch velocity of each player's home runs
sns.boxplot(x='player_name', y='release_speed', color='tab:blue', data=judge_stanton_h
```

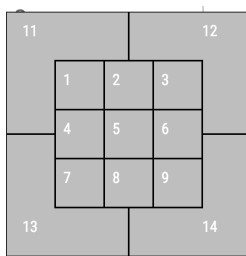
Out Text(0.5, 1.0, 'Home Runs, 2015-2017')

[23]:



6. Home runs by pitch location (I)

So Judge appears to hit his home runs off of faster pitches than Stanton. We might call Judge a fastball hitter. Stanton appears agnostic to pitch speed and likely pitch movement since slower pitches (e.g. curveballs, sliders, and changeups) tend to have more break. Statcast *does* track pitch movement and type but let's move on to something else: **pitch location**. Statcast tracks the zone the pitch is in when it crosses the plate. The zone numbering looks like this (from the catcher's point of view):



We can plot this using a 2D histogram. For simplicity, let's only look at strikes, which gives us a 9x9 grid. We can view each zone as coordinates on a 2D plot, the bottom left corner being (1,1) and the top right corner being (3,3). Let's set up a function to assign x-coordinates to each pitch.

```
In [24]: def assign_x_coord(row):
    """
    Assigns an x-coordinate to Statcast's strike zone numbers. Zones 11, 12, 13,
    and 14 are ignored for plotting simplicity.
    """
    # Left third of strike zone
    if row.zone in [1, 4, 7]:
        return 1
    # Middle third of strike zone
    if row.zone in [2, 5, 8]:
        return 2
    # Right third of strike zone
    if row.zone in [3, 6, 9]:
        return 3
```

7. Home runs by pitch location (II)

And let's do the same but for y-coordinates.

```
In [25]: def assign_y_coord(row):
    """
    Assigns a y-coordinate to Statcast's strike zone numbers. Zones 11, 12, 13,
    and 14 are ignored for plotting simplicity.
    """
    # Upper third of strike zone
    if row.zone in [1, 2, 3]:
        return 3
    # Middle third of strike zone
    if row.zone in [4, 5, 6]:
        return 2
    # Lower third of strike zone
    if row.zone in [7, 8, 9]:
        return 1
```

8. Aaron Judge's home run zone

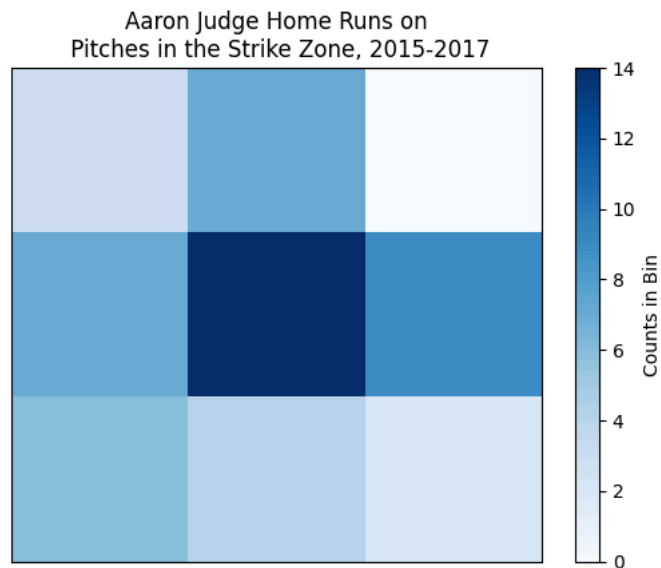
Now we can apply the functions we've created then construct our 2D histograms. First, for Aaron Judge (again, for pitches in the strike zone that resulted in home runs).

```
In [26]: # Zones 11, 12, 13, and 14 are to be ignored for plotting simplicity
judge_strike_hr = judge_hr.copy().loc[judge_hr.zone <= 9]

# Assign Cartesian coordinates to pitches in the strike zone for Judge home runs
judge_strike_hr['zone_x'] = judge_strike_hr.apply(assign_x_coord, axis=1)
judge_strike_hr['zone_y'] = judge_strike_hr.apply(assign_y_coord, axis=1)

# Plot Judge's home run zone as a 2D histogram with a colorbar
plt.hist2d(judge_strike_hr['zone_x'], judge_strike_hr['zone_y'], bins = 3, cmap='Blue')
```

```
plt.title('Aaron Judge Home Runs on\n Pitches in the Strike Zone, 2015-2017')
plt.gca().get_xaxis().set_visible(False)
plt.gca().get_yaxis().set_visible(False)
cb = plt.colorbar()
cb.set_label('Counts in Bin')
```



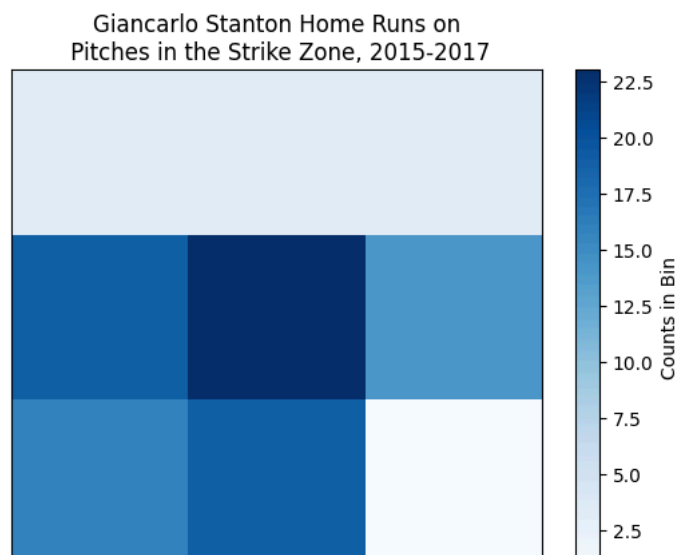
9. Giancarlo Stanton's home run zone

And now for Giancarlo Stanton.

```
In [27]: # Zones 11, 12, 13, and 14 are to be ignored for plotting simplicity
stanton_strike_hr = stanton_hr.copy().loc[stanton_hr.zone <= 9]

# Assign Cartesian coordinates to pitches in the strike zone for Stanton home runs
stanton_strike_hr['zone_x'] = stanton_strike_hr.apply(assign_x_coord, axis=1)
stanton_strike_hr['zone_y'] = stanton_strike_hr.apply(assign_y_coord, axis=1)

# Plot Stanton's home run zone as a 2D histogram with a colorbar
plt.hist2d(stanton_strike_hr['zone_x'], stanton_strike_hr['zone_y'], bins = 3, cmap='
plt.title('Giancarlo Stanton Home Runs on\n Pitches in the Strike Zone, 2015-2017')
plt.gca().get_xaxis().set_visible(False)
plt.gca().get_yaxis().set_visible(False)
cb = plt.colorbar()
cb.set_label('Counts in Bin')
```



10. Should opposing pitchers be scared?

A few takeaways:

- Stanton does not hit many home runs on pitches in the upper third of the strike zone.
- Like pretty much every hitter ever, both players love pitches in the horizontal and vertical middle of the plate.
- Judge's least favorite home run pitch appears to be high-away while Stanton's appears to be low-away.
- If we were to describe Stanton's home run zone, it'd be middle-inside. Judge's home run zone is much more spread out.

The grand takeaway from this whole exercise: Aaron Judge and Giancarlo Stanton are not identical despite their superficial similarities. In terms of home runs, their launch profiles, as well as their pitch speed and location preferences, are different.

Should opposing pitchers still be scared?

```
In [28]: # Should opposing pitchers be wary of Aaron Judge and Giancarlo Stanton  
         should_pitchers_be_scared = True
```