# Manipulating data tables II



### Overview

Lab 2 discussion

Discussion of chapter 2 of Astroball

Review of data manipulation methods and steps

#### Working together on:

- Warm up exercises
- Calculating the run expectancy matrix

#### Lab 3

Calculate the probability a team will win a game given the inning/outs and run differential

## Lab 2: questions?

How did it go?

### Astroball discussion

Let's discuss the chapter for 7 minutes in breakout rooms and then have a larger conversation as a group

• Discuss your quote and reaction to chapter 2

Thoughts on chapter 2 of Astroball?

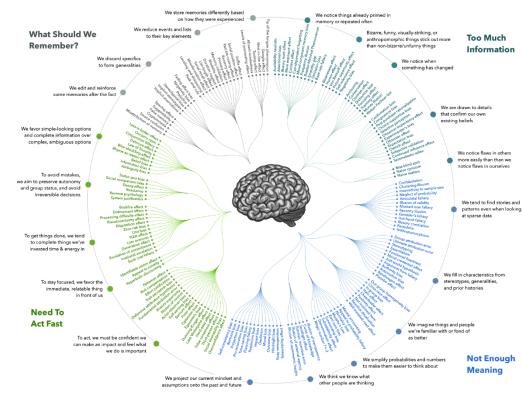
### Astroball: making the best decisions

Using linear regression to combining performance statistics and scouting reports.

### Kahneman and Tversky

- Behavioral economics vs. assuming rational agent.
- Cognitive biases

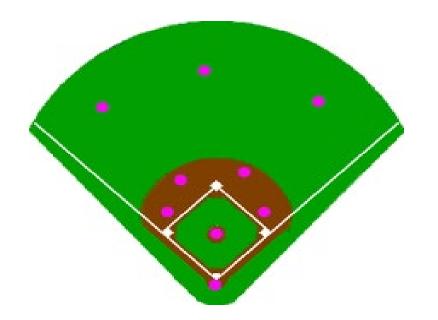
#### COGNITIVE BIAS CODEX



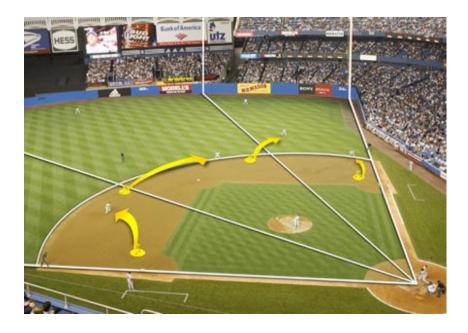
DESIGNHACKS.CO - CATEGORIZATION BY BUSTER BENSON - ALGORITHMIC DESIGN BY JOHN MANOOGIAN III (JM3) - DATA BY WIKIPEDIA

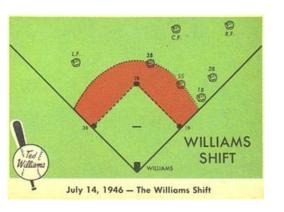
### Astroball: infield shifts

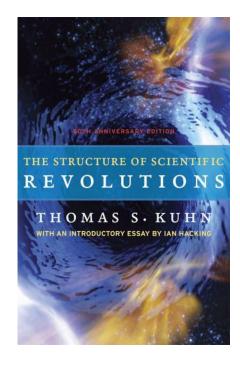
#### Normal positioning



#### Defensive shift







### Astroball: the butt slide

Deadspin from every angle

MLB analysis

### Data manipulation part II

Let's continue getting practice with data manipulation/wrangling by working in groups to do more advanced exercises.

- Quick review of key datascience method and data manipulation steps
- Warm up exercises: examining batting order and lefty/righty matchups
- Run expectancy matrix
- Lab 3: probability of winning a game given the current situation

Very quick review of datascience methods

# Selecting and filtering

Description	datascience package	Pandas
Select a single column as a Table/Series	tb.select("col1")	df.col1 also df["col1"]
Select a single column as a ndarray	tb["col1"]	df["col1"].to_numpy()
Select multiple columns as a Table/DataFrame	tb.select("col1", "col2")	df[["col1", "col2"]]
Select 20th row	tb.take(20)	df.iloc[[20]]
Filtering rows equal to cond	tb.where("col", cond)	df[df.col == cond]
Sort descending by column	tb.sort("col", descending = True)	df.sort_values("col", ascending = False)

### Data aggregation

Data aggregation/summarization consists of reducing a data table to a summary, often conditioned on a groping variable.

Q: Have we seen a data aggregation function yet?

A: tb.group('grouping\_var', aggregation\_function)

- tb.group('age') # counts how many players are a particular age
- tb.group('yearID', max) # get the max statistic value for each year

#### **Pandas**

df.groupby('grouping\_var').aggfunc().reset\_index()

If this isn't used then the grouping column becomes the Index

### Data aggregation

Other datascience aggregation functions:

tb.groups(['col1', 'col2'], aggregation function)

tb.pivot('col\_var', 'row\_var', values = 'fill\_var', collect = agg\_function)

>>> t:	itanic		
age	survival	gender	prediction
21	0	M	0
44	0	M	0
56	0	M	1
89	1	M	1
95	1	F	0
40	1	F	1
80	0	F	0
45	1	F	1
		٠.	

"long format"

"wide format"

### Data aggregation

#### **Pandas**

>>> titanic			
age	survival	gender	prediction
21	0	M	0
44	0	M	0
56	0	M	1
89	1	M	1
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40	1	F	1
80	0	F	0
45	1	F	1

### Numpy operations

#### Combining ndarrays

combo\_ndarray = ndarray1.append(ndarray2)

#### Casting: converts from one type to another type

- ndarray.astype('int') # could be useful to convert Booleans to numbers
- ndarray.astype('str') # could be useful to convert numbers to characters

#### Combining strings in ndarrays

- np.char.add(strings\_ndarray, '---', )
- np.char.add(strings\_ndarray1, strings\_ndarray2)

# Data processing steps

### Data processing steps

- 1. Data cleaning: getting data into a table-like format (tidy)
  - 80% of Data Scientists time is spent cleaning data
  - 20% of Data Scientists time is spent complaining about cleaning data
- 2. Data manipulation: get columns you need for subsequent analyses
  - Creating new columns from existing columns
  - Joining data tables
  - Etc.
- 3. Data aggregation: extract key summary statistics from data
- 4. Visualize: create clear figures to reveal insights

### Thinking through data manipulation steps

- 2. Data manipulation: get columns you need for subsequent analyses
- 3. Data aggregation: extract key summary statistics from data

To think through data manipulation steps it is useful to:

- 1. Think about final aggregated result one wants
- 2. Think about the data table (i.e., columns needed) that could be aggregated to get the final result
  - Second to last step
- 3. Think about how to get data table in step 2 from starting data table

# Data manipulation exercises

It is widely believed that the best home run hitters usually bat third or fourth in the line up.

Let's check to see if this is the case!

Let's use the play-by-play retro data from 2019



Let's start with the retrosheet play-by-play data and go through these steps:

- 1. Think about final aggregated result one wants, what would it look like?
- 2. Think about the data table (i.e., columns needed) that could be aggregated to get the final result
  - Second to last step
- 3. Think about how to get data table in step 2 from starting data table

Let's start with the retrosheet playby-play data

- 1. Think about final aggregated result one wants, what would it look like?
  - i.e., what are the variables in the final aggregated table?
  - Sketch it

HR frequency

These values filled in

Let's start with the retrosheet playby-play data

- 2. Think about the data table (i.e., columns needed) that could be aggregated to get the final result
  - Second to last step

What variables are needed?

Batting order	HR frequency
1	
2	
3	
4	
5	
6	
7	
8	
9	

Let's start with the retrosheet playby-play data

3. Think about how to get data table in step 2 from starting data table

Batting order	HR frequency
1	
2	
3	
4	
5	
6	
7	
8	
9	

Let's go into breakout rooms and solve these together...

Get in class\_04 Jupyter notebook

Pandas people together:

• Gabby, Harry, Teddy, anyone else?

### Righties vs. lefties

1. Do righthanded batters hit more home runs than lefthanded batters?

- 2. Do right/lefthanded batters hit more home runs off right/lefthanded pitchers?
  - A platoon is a method of sharing playing time, where two players are selected to play a single defensive position. Usually, one platoon player is right-handed and the other is left-handed.



### Run expectancy matrix

The run expectancy matrix contains the expected number of runs given the number of outs, and base runners

- E[ P(R|base runners, outs)]
- History:
  - First described in 1963 by George Lindsey in an article in Operations Research
  - Discussed extensively in The Book by Tango, Lichtman and Dolphin

Q: How many out-runner game states are there in half an inning?

• A: 24

#### Empty run expectancy matrix

Bases	0 Outs	1 Out	2 Outs
3			
_ 2 _			
_ 2 3			
1			
1_3			
12_			
123			

### Run expectancy matrix

The **run expectancy matrix** contains the expected number of runs given the number of outs, and base runners

Let's try to fill in the run expectancy matrix with your own best guesses

- Go to: bit.ly/runexpect
- Copy over the template to your own sheet
- Fill in with your best guess

Empty run expectancy matrix

Bases	0 Outs	1 Out	2 Outs
3			
_ 2 _			
_ 2 3			
1			
1_3			
12_			
123			

- **2. Data manipulation**: get columns you need for subsequent analyses
- 3. Data aggregation: extract key summary statistics from data

To think through data manipulation steps it is useful to:

- 1. Think about final aggregated result one wants
- 2. Think about the data table (i.e., columns needed) that could be aggregated to get the final result
  - Second to last step
- 3. Think about how to get data table in step 2 from starting data table

Do part 2.1.1, 2.1.2, 2.1.3 in a breakout room and then let's discuss these steps as a class

Bases	0 Outs	1 Out	2 Outs
3			
_ 2 _			
_ 2 3			
1			
1_3			
12_			
123			

Question 2.1.1 - Final goal: What are the...?

- 1. rows: runners on base (8 states)
- 2. columns: outs (3 states)
- **3. fill values:** expected number of runs scored by the end of the inning given the current game state.

Bases	0 Outs	1 Out	2 Outs
3			
_ 2 _			
_ 2 3			
1			
1_3			
12_			
123			

Question 2.1.2 - The aggregating methods are:

- pivot: Generate a table with a column for each unique value in columns, with rows for each unique value in rows.
- **2. group**: Group rows by unique values in a column; count or aggregate others.
- **3. groups**: Group rows by multiple columns, count or aggregate others.

Question 2.1.3 - The columns that are needed in the $2^{nd}$ to last step are:		BASE_RUNNER_STATE	оитѕ_ст	RUN_	ROI
1.	The state of the base runners	000		0	0
	• e.g., bases empty, runner on first, etc.	001		0	0
2.	The number of outs	001		1	0
3.	The number of runs scored that remain to be	011		1	0
	scored in the inning (i.e., the final score in the inning minus the current score).	000		1	0
	w would you aggregate these results to get e final run expectancy matrix?				

Let's work in groups to try to calculate the run expectancy matrix

### Lab 3

Calculate the probability of winning a game given the inning and score difference

Might find it challenging so start early!!!

• Use Ed discussions for questions and answer other's questions