# 9) Sampling and Empirical Distributions

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#### Reference

Tables, Graphics, and Figures from

### Computational and Inferential Thinking: The Foundations of Data Science

Adhikari & DeNero (2019): Ch 10 Sampling and Empirical Distributions

https://www.inferentialthinking.com/

#### **Top Movies Dataset**

```
import numpy as np
from datascience import *
path_data = 'https://github.com/data-8/textbook/raw/gh-pages/data/'
top1 = Table.read_table(path_data + 'top_movies.csv')
top2 = top1.with_column('Row Index', np.arange(top1.num_rows))
top = top2.move_to_start('Row Index')
top.set_format(make_array(3, 4), NumberFormatter)
```

Row Index	Title	Studio	Gross
0	Star Wars: The Force Awakens	Buena Vista (Disney)	906,723,418
1	Avatar	Fox	760,507,625
2	Titanic	Paramount	658,672,302

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#### **Deterministic Samples**

top.take(make\_array(3, 18, 100))

Row Index	Title	Studio	Gross
3	Jurassic World	Universal	652,270,625
18	Spider-Man	Sony	403,706,375
100	Gone with the Wind	MGM	198,676,459

top.where('Title', are.containing('Harry Potter'))

Row Index	Title	Studio	Gross
22	Harry Potter and the Deathly Hallows Part 2	Warner Bros.	381,011,219
43	Harry Potter and the Sorcerer's Stone	Warner Bros.	317,575,550

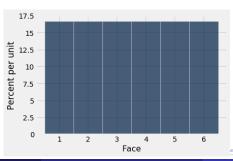
#### **Systematic Sample**

#### Random start among rows 0 through 9

```
start = np.random.choice(np.arange(10))
top.take(np.arange(start, top.num_rows, 10))
```

Studio	Title	Row Index
Paramount	Titanic	2
Lionsgate	The Hunger Games: Catching Fire	12
Warner Bros.	Harry Potter and the Deathly Hallows Part 2	22

#### **Probability Distribution**

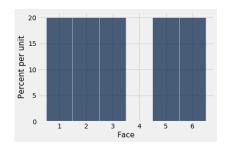


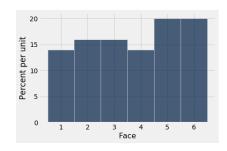
#### **Empirical Distributions**

```
def empirical_hist_die(n):
    die.sample(n).hist(bins = die_bins)
```

empirical\_hist\_die(10)

empirical\_hist\_die(100)





#### The Law of Averages

In the long run, the empirical probability gets closer and closer to the theoretical probability of the event

empirical\_hist\_die(1000)



$$\left(\frac{1}{6}\cong 0.17\right)$$

Independently and under identical conditions: every repetition is performed in the same way regardless of the results of all the other repetitions

#### **Bureau of Transportation Statistics in US**

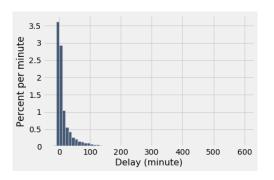
## United Airlines domestic flights departing from San Francisco in Summer of 2015

```
united = Table.read_table(path_data + 'united_summer2015.csv')
```

Date	Flight Number	Destination	Delay
6/1/15	73	HNL	257
6/1/15	217	EWR	28
6/1/15	237	STL	-3

#### Normal Distribution

```
delay_bins = np.append(np.arange(-20, 301, 10), 600)
united.hist('Delay', bins = delay_bins, unit = 'minute')
```

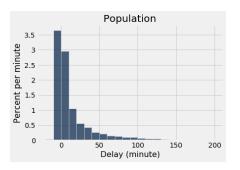


united.where('Delay', are.above(200)).num\_rows/united.num\_rows

0.008390596745027125

#### Ignore the 0.8% of flights

```
delay_bins = np.arange(-20, 201, 10)
united.hist('Delay', bins = delay_bins, unit = 'minute')
plots.title('Population');
```

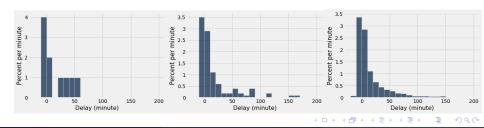


0.2935985533453888



#### Random Sample converges to the Population

```
def empirical_hist_delay(n):
    united.sample(n).hist('Delay',
    bins = delay_bins, unit = 'minute')
empirical_hist_delay(10)
empirical_hist_delay(100)
empirical_hist_delay(1000)
```



#### Parameter vs Statistic

```
np.median(united.column('Delay'))
                                      2.0
 united.where('Delay',
   are.below or equal to(2)).num rows / united.num rows
               0.5018444846292948
united.where('Delay', are.equal to(2)).num rows
                        480
np.median(sample 1000.column('Delay'))
                                         1.0
np.median(united.sample(1000).column('Delay'))
                        2.5
```

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#### Simulating a Statistic 5,000

```
def random_sample_median():
    return np.median(united.sample(1000).column('Delay'))
medians = make_array()

for i in np.arange(5000):
    medians = np.append(medians, random_sample_median())

simulated_medians = Table().with_column('Sample Median', medians)
simulated medians.hist(bins=np.arange(0.5, 5, 1))
```

