## **Sampling Distributions Introduction**

In order to gain a bit more comfort with this idea of sampling distributions, let's do some practice in python.

Below is an array that represents the students we saw in the previous videos, where 1 represents the students that drink coffee, and 0 represents the students that do not drink coffee.

1. Find the proportion of students who drink coffee in the above array. Store this value in a variable **p**.

```
In [2]: students.mean()
Out[2]: 0.7142857142857143
In [3]: p = students.mean()
p
Out[3]: 0.7142857142857143
```

2. Use numpy's **random.choice** to simulate 5 draws from the students array. What is proportion of your sample drink coffee?

3. Repeat the above to obtain 10,000 additional proportions, where each sample was of size 5. Store these in a variable called <code>sample\_props</code>.

```
In [6]: sample_props = []
for _ in range(10000):
    sample = np.random.choice(students, 5, replace=True)
    sample_props.append(sample.mean())
```

4. What is the mean proportion of all 10,000 of these proportions? This is often called **the mean** of the sampling distribution.

```
In [7]: sample_props = np.array(sample_props)
sample_props.mean()
```

- Out[7]: 0.7139999999999997
  - 5. What are the variance and standard deviation for the original 21 data values?
- In [8]: print('The standard deviation for the original data is {}'.format(students.
  print('The variance for the original data is {}'.format(students.var()))

The standard deviation for the original data is 0.45175395145262565 The variance for the original data is 0.20408163265306126

- 6. What are the variance and standard deviation for the 10,000 proportions you created?
- In [9]: print('The standard deviation of the sampling distribution of the mean of 5
  print('The variance for the sampling distribution of the mean of 5 draws is

The standard deviation of the sampling distribution of the mean of 5 draw s is 0.2043624231604235The variance for the sampling distribution of the mean of 5 draws is 0.04

7. Compute p(1-p), which of your answers does this most closely match?

```
In [10]: p*(1-p) # The variance of the original data
```

Out[10]: 0.20408163265306123

176399999999996

8. Compute p(1-p)/n, which of your answers does this most closely match?

```
In [11]: p*(1-p)/5 # The variance of the sample mean of size 5
```

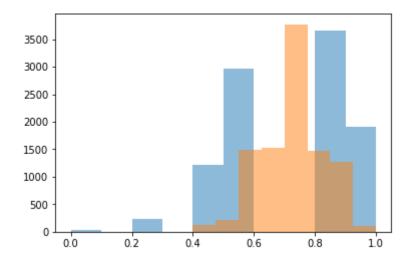
- Out[11]: 0.040816326530612249
  - 9. Notice that your answer to 8. is commonly called the **variance of the sampling distribution**. If you were to change your first sample to be 20, what would this do for the variance of the sampling distribution? Simulate and calculate the new answers in 6. and 8. to check that the consistency you found before still holds.

```
In [12]: ##Simulate your 20 draws
sample_props_20 = []
for _ in range(10000):
    sample = np.random.choice(students, 20, replace=True)
    sample_props_20.append(sample.mean())
```

0.010300994375

10. Finally, plot a histgram of the 10,000 draws from both the proportions with a sample size of 5 and the proportions with a sample size of 20. Each of these distributions is a sampling distribution. One is for the proportions of sample size 5 and the other a sampling distribution for proportions with sample size 20.

```
In [14]: plt.hist(sample_props, alpha=.5);
plt.hist(np.array(sample_props_20), alpha=.5);
```



```
In [15]: # Notice the 20 is much more normally distributed than the 5

In []:
```

https://viewa955e67f.udacity-student-workspaces.com/notebooks/Sampling Distributions-Solution.ipynb