

## flip function

```
# generate a 'flip'
def flip(num = 1):
    flips = []

    for i in range(num):
        num = np.random.uniform(low=0.0, high=1.0)
        if num > 0.5:
            flips.append('H')
        else:
            flips.append('T')
    return flips
```

*# should be doing yield here if you know 'generators'*

```
5]: flips = flip(10)
    print(flips)
```

```
['T', 'T', 'T', 'T', 'H', 'T', 'H', 'H', 'H', 'H']
```

```
] : values, counts = np.unique(flips, return_counts=True)
```

you need to reproduce the problem and experiment  
do not hard code made the code predicatble

```
def flip(num = 1):
    flips = []

    for i in range(num):
        num = np.random.uniform(low=0.0, high=1.0)
        if num > 0.5:
            flips.append('H')          # should be doing yield here if you know 'generators'
        else:
            flips.append('T')
    return flips

# Flip
flips = flip(10)
values, counts = np.unique(flips, return_counts=True)

# print values/stats
# print(flip())
print(flips)
print(counts)
```

```
['H', 'T', 'H', 'H', 'H', 'T', 'T', 'H', 'H', 'H']
[7 3]
```

```
['H', 'T', 'H', 'H', 'H', 'T', 'T', 'H', 'H', 'H']
[7 3]
```

# Heisenbugs

## Probability of Flips

# Reproducible Randomness

Seeds not to take the real world instead take from the zero

```
] : # computers are 'deterministic'. You can not do 'random' in computers!  
# So, you start with some 'seed' then do deterministic things  
# this is called pseudo-randomness  
  
# sometimes you want to suppress this!  
  
np.random.seed(0) # random numbers and seed
```

must import seed before start the experiment

```
] import matplotlib
import matplotlib.pyplot as plt
%matplotlib inline

import numpy as np

import seaborn as sns
sns.set(color_codes=True)
sns.set_style("white")    # See more styling options here: https://seaborn.pydata.org/tutorial/

# np.random.seed(0)    # random numbers and seed

# generate a 'flip'
def flip(num = 1):
    flips = []

    for i in range(num):
        num = np.random.uniform(low=0.0, high=1.0)
        if num > 0.5:
            flips.append('H')    # should be doing yield here if you know 'generators'
        else:
            flips.append('T')
    return flips

# Flin
```

## Probability of Flips

```
[ ]: from collections import Counter, defaultdict

def get_freqs(flips):
    keys = Counter(flips).keys()
    vals = Counter(flips).values()

    # print(keys)
    # print(vals)

    # return dict(zip(keys, vals))    # bug: what if there are no 'H' or no 'T'

    return defaultdict(int, dict(zip(keys, vals)))
```

```
def get_freqs(flips):  
    keys = Counter(flips).keys()  
    vals = Counter(flips).values()  
  
    # print(keys)  
    # print(vals)  
  
    # return dict(zip(keys, vals))    # bug: what if there are no 'H' or no 'T'  
  
    return defaultdict(int, dict(zip(keys, vals)))
```


making Dictionary from the counter of values  
problem what happen there is not tail that leads to key  
error.

We will use the default dictionary that not lead to the  
key error.

## Probability of Flips

```
] from collections import Counter, defaultdict
```

```
def get_freqs(flips):  
    keys = Counter(flips).keys()  
    vals = Counter(flips).values()  
  
    # print(keys)  
    # print(vals)  
  
    # return dict(zip(keys, vals))    # bug: what if there are no 'H' or no 'T'  
  
    return defaultdict(int, dict(zip(keys, vals)))
```





## Experiment: Prob calculated based on 1 flip upto N flips

```
[ ]: maximum_flips = 1000

probs = []
for num_flips in range(1, maximum_flips):
    flips = flip(num_flips)
    freqs = get_freqs(flips)
    prob_h = freqs['H'] / len(flips)

    probs.append(prob_h)

# print(probs)
```

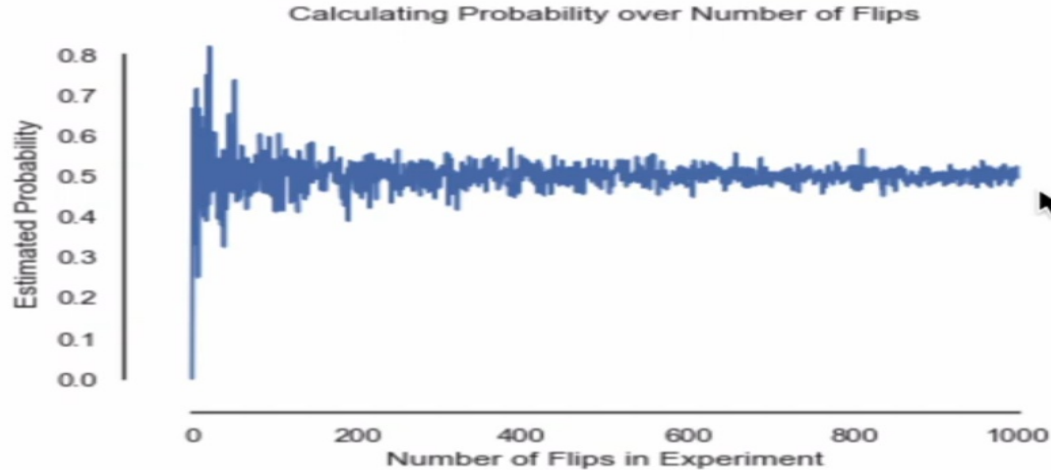
```
[ ]: print(freqs)
```

i tose the coin  
1000 times then  
probaility of  
head and tail in  
represented in  
the graph

Tas 1 get head  
Tas 2 get head two time  
Tas 3 get head 3 times

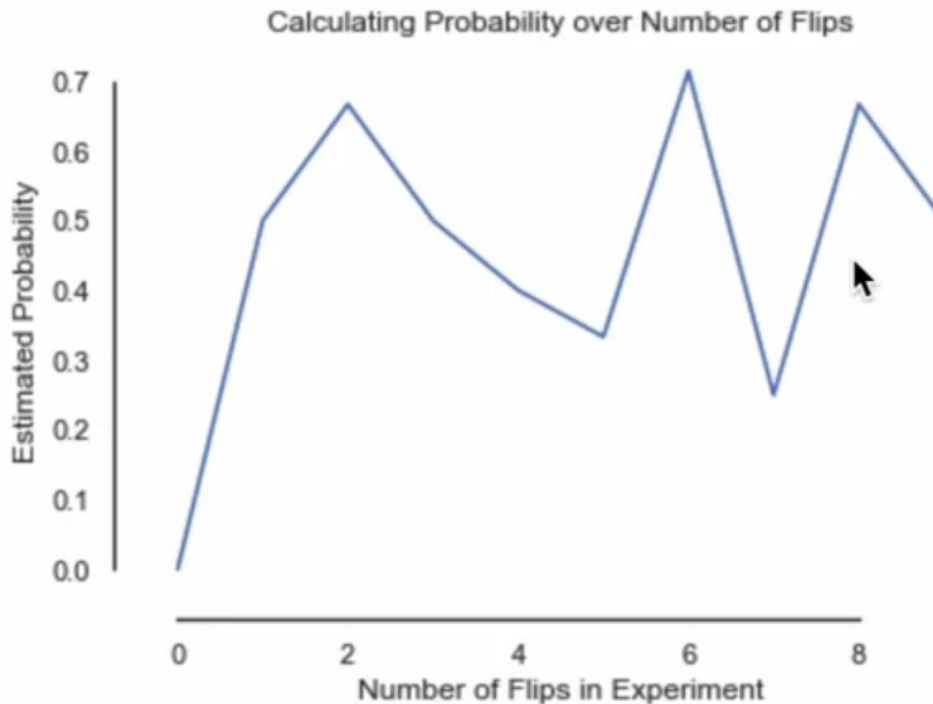
using this experiment we  
have proof the proabilty of  
head more tne tail

```
[84]: plt.plot(probs)
plt.ylabel('Estimated Probability')
plt.xlabel('Number of Flips in Experiment');
plt.title("Calculating Probability over Number of Flips")
sns.despine(offset=10, trim=True); # move axes away
plt.show()
```



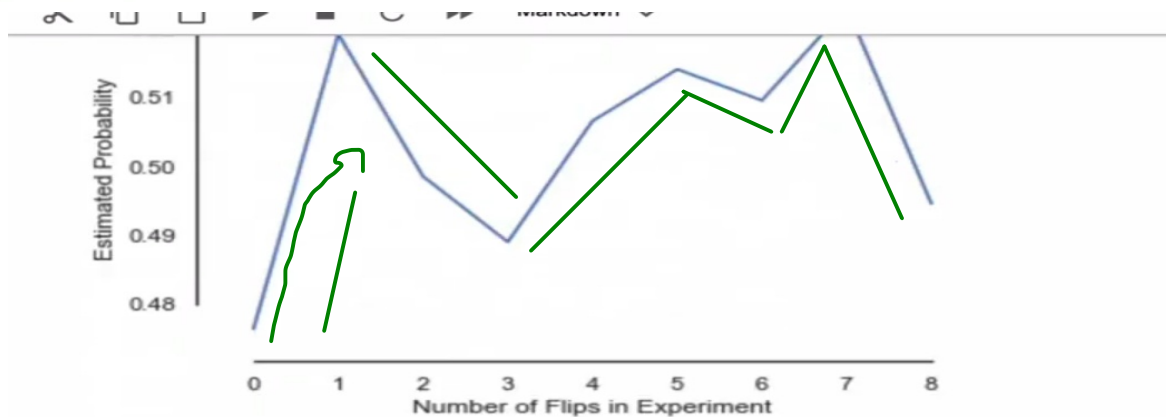
The probability of last 10 value  
form the prabaility

```
[83]: plt.plot(probs[:10])
plt.ylabel('Estimated Probability')
plt.xlabel('Number of Flips in Experiment');
plt.title("Calculating Probability over Number of Flips")
sns.despine(offset=10, trim=True); # move axes away
plt.show()
```



Frist Values of probaility of  
the graph

```
] : plt.plot(probs[maximum_flips-10:])  
plt.ylabel('Estimated Probability')  
plt.xlabel('Number of Flips in Experiment');  
plt.title("Calculating Probability over Number of Flips")  
sns.despine(offset=10, trim=True); # move axes away  
plt.show()
```



this is library for interactive the graph more efficiently

## Bokeh For Interactive Plots

```
[ ]: !pip install bokeh
```

```
[ ]: from bokeh.io import show, output_notebook
     from bokeh.plotting import figure

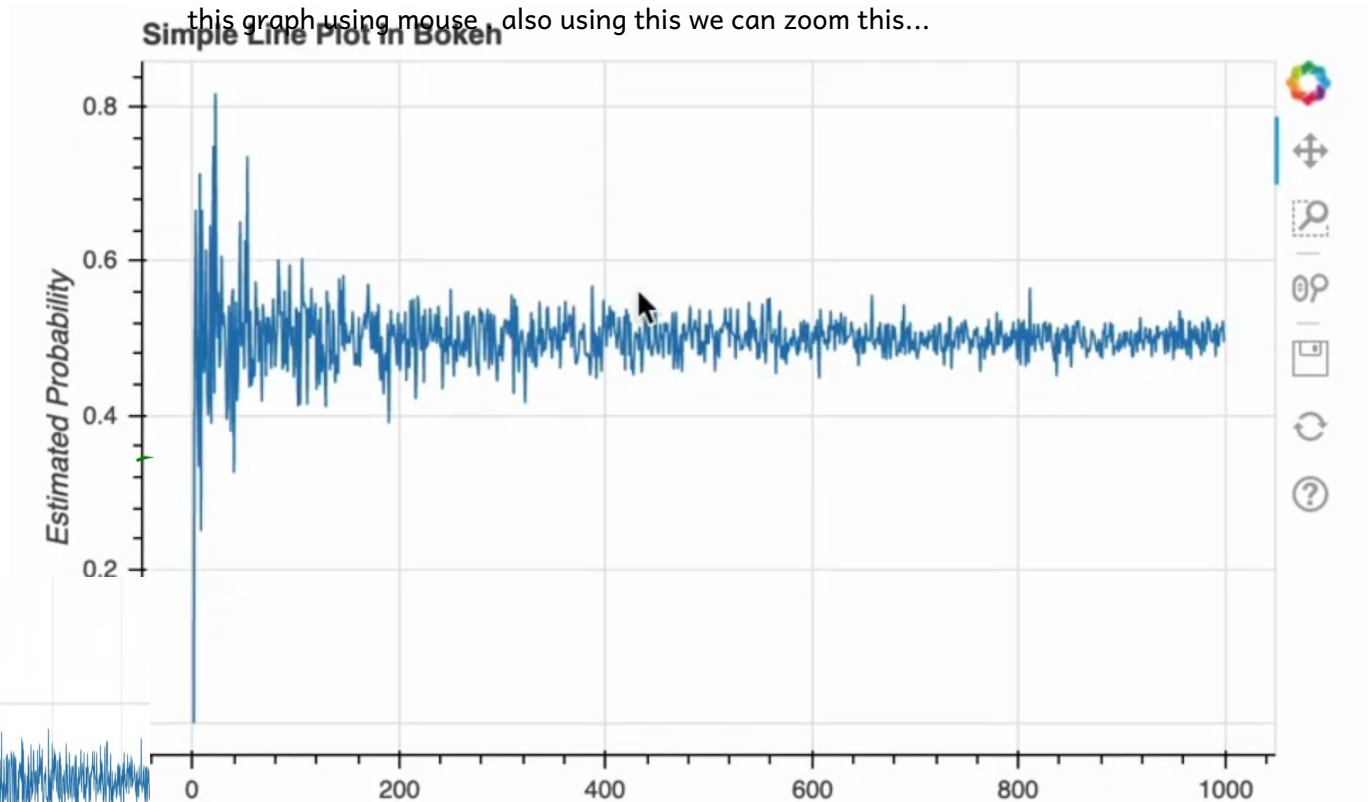
     output_notebook()
```

```
[ ]: p = figure(title="Simple Line Plot in Bokeh",
                x_axis_label='Number of Flips in Experiment',
                y_axis_label='Estimated Probability',
                plot_width=580, plot_height=380)
```

```
[ ]: # Add a line renderer with legend and line thickness
     x = range(1, maximum_flips)
     p.line(x=x, y=probs)

     # Show the results
     show(p)
```

This is output of graph of bokey library whic i told the most efficeintly way of representation of graph .you can also pan this graph using mouse also using this we can zoom this...



Most important Question :

what is chance of head in one flip?

what is chance of head in two flip?

what is chance of head in 100 flip?

what is chance of head in 1000 flip?

Then we do the experiment and calculate one times two times upto 1000 times.

But also we can't experiment more then one times let Election we want to predict who will win in this case you can't do the experiemnt you have to predict from the previos knowledge . Other example Who will the Tournament?

## Quantifying Chances

- Flip a coin:

→ hard

Q: "What are the chances that it will land on its head?"

Not: "If I flip it 10 times, how many will be heads?"

→ easy