```
In [1]: import time
   import numpy as np
   from scipy.stats import norm, beta
```

Question 0 - Markdown warmup

This is *question 0* for <u>problem set 1 (https://jbhender.github.io/Stats507/F21/ps/ps1.html)</u> of <u>Stats 507 (https://jbhender.github.io/Stats507/F21/)</u>.

```
Question 0 is about Markdown.
```

The next question is about the **Fibonnaci sequence**, $F_n = F_{n-2} + F_{n-1}$. In part **a** we will define a Python function fib_rec().

Below is a ...

Level 3 Header

Next, we can make a bulleted list:

- Item 1
 - detail 1
 - detail 2
- Item 2

Finally, we can make an enumerated list:

- 1. Item 1
- 2. Item 2
- 3. Item 3

Question 1 - Fibonnaci Sequence

a.

```
Returns
-----
the nth Fibonacci numbers.

if n == 0:
    return a
elif n == 1:
    return b
else:
    return fib_rec(n-1) + fib_rec(n-2)
```

```
In [3]: print(fib_rec(7))
print(fib_rec(11))
print(fib_rec(13))

13
89
233
```

```
In [4]: def fib_for(n, a = 0, b = 1):
            This function generates Fibonacci numbers by using for loop.
            Parameters
            n : int
                the nth Fibonacci numbers will be calculated.
            a : int, optional
                the first Fibonacci numbers F_0. The default is 0.
            b : int, optional
                the second Fibonacci numbers F_1. The default is 1.
            Returns
            the nth Fibonacci numbers.
            .....
            x = a
            y = b
            for k in range(n):
                z = x + y
                x = y
                y = z
            return x
```

b.

```
In [5]: print(fib_for(7))
    print(fib_for(11))
    print(fib_for(13))

    13
    89
    233
c.
```

```
In [6]: def fib_whl(n, a = 0, b = 1):
            This function generates Fibonacci numbers by using while loop.
            Parameters
            n : int
                the nth Fibonacci numbers will be calculated.
            a : int, optional
                 the first Fibonacci numbers F_0. The default is 0.
            b : int, optional
                 the second Fibonacci numbers F_1. The default is 1.
            Returns
            the nth Fibonacci numbers.
            0.00
            x = a
            y = b
            k = 0
            while k < n:</pre>
                z = x + y
                x = y
                y = z
                 k += 1
             return x
```

```
In [7]: print(fib_whl(7))
print(fib_whl(11))
print(fib_whl(13))

13
89
233
```

.

d.

```
In [8]: def fib_rnd(n):
             This function generates Fibonacci numbers by using the rounding me
             Parameters
                 the nth Fibonacci numbers will be calculated.
             Returns
             the nth Fibonacci numbers.
             phi = (1 + 5 ** .5) / 2
             return round(phi ** n / 5 ** .5)
In [9]: print(fib_rnd(7))
         print(fib_rnd(11))
         print(fib_rnd(13))
         13
         89
         233
         e.
In [10]: def fib_flr(n, a = 0, b = 1):
             This function generates Fibonacci numbers by using the truncation
             Parameters
             n : int
                 the nth Fibonacci numbers will be calculated.
         in
             Returns
             the nth Fibonacci numbers.
             phi = (1 + 5 ** .5) / 2
             return int(phi ** n / 5 ** .5 + .5)
```

```
print(fib_flr(11))
         print(fib_flr(13))
         13
         89
         233
         f.
In [12]: def calculate_time(func, n, r):
             This function calculates the median running time of each functions
             Parameters
             func: function
                 the function name.
             n : int
                 the nth Fibonacci numbers will be calculated.
             r : int
                 repeated number of calculating the running time.
              Returns
              the median running time of each functions.
             cal_time = []
             for i in range(r):
                 start = time.perf_counter()
                 func(n)
                 end = time.perf_counter()
                 cal time.append(end - start)
             return round(np.mean(cal_time), 4)
```

n	fib_rec	fib_for	fib_whl	fib_rnd	fib_flr
15	0.0002	0.0	0.0	0.0	0.0
20	0.0024	0.0	0.0	0.0	0.0
25	0.0261	0.0	0.0	0.0	0.0
30	0.2826	0.0	0.0	0.0	0.0
35	3.1392	0.0	0.0	0.0	0.0

Question 2 - Pascal's Triangle

a.

In [11]: print(fib_flr(7))

```
n: int
    compute the nth row of Pascal's triangle.

Returns
----
the nth row of Pascal's triangle.

x = [1]
for i in range(n - 1):
    c_i = int(x[i] * (n - i - 1) / (i + 1))
    x.append(c_i)
return x
```

b.

```
In [14]: def print_pascal_trl(n):
    """
    This function prints the first n rows of Pascal's triangle.

Parameters
------
n: int
    print the first n rows of Pascal's triangle.

Returns
-----
None.

"""
largest_elm = pascal_row(n)[n // 2]
elm_width = len(str(largest_elm)) + 1
trl_width = elm_width * n

for i in range(n):
    x = pascal_row(i + 1)
    y = "".join([str(x[j]).center(elm_width) for j in range(i + 1)
    print(y.center(trl_width))
```

Tn [15]. | nmint naccal tn]/11\

In [15]: |print_pascal_trl(11)

```
1
1 1
1 2 1
1 3 3 1
1 4 6 4 1
1 5 10 10 5 1
1 6 15 20 15 6 1
1 7 21 35 35 21 7 1
1 8 28 56 70 56 28 8 1
1 9 36 84 126 126 84 36 9 1
1 10 45 120 210 252 210 120 45 10 1
```

Question 3 - Statistics 101

a.

```
In [16]: def estimate_ci(data, level, output = "string"):
             This function estimates a point and interval based on Normal theor
             Parameters
             data : array
                 input data.
             level : int
                 confidence interval.
             output : string
             Returns
             string or dictionary.
             mean = np.mean(data)
             std = np.std(data)
             z = norm.ppf(1 - (1 - level * .01) / 2) / (len(data) ** .5)
             est = round(mean, 3)
             lwr = round(mean - z * std, 3)
             upr = round(mean + z * std, 3)
             if output == "string":
                 return str(est) + '[' + str(level) + '%CI:(' + str(lwr) + ','
             elif output == None:
                 return {'est': est, 'lwr': lwr, 'upr': upr}
```

b.

```
Parameters
data: array
    input data.
method : string
    indicate which method will be used.
    "normal_appx" means normal approximation will be used.
    "clopper_pearson" means Clopper-Pearson interval will be compu
    "Jeffrey" means Jeffrey's interval interval will be computed.
    "Agresti-Coull" means Agresti-Coull interval will be computed.
level : int
    confidence interval.
output : string
Returns
string or dictionary.
x = np.sum(data)
n = len(data)
p_hat = x / n
alpha = 1 - level * .01
if method == "normal appx":
    est = round(p_hat, 3)
    if np.min([n * p hat, n * (1 - p hat)]) > 12:
        z = norm.ppf(1 - alpha / 2)
        lwr = round(p_hat - z * ((p_hat * (1 - p_hat) / n) ** .5),
        upr = round(p_hat + z * ((p_hat * (1 - p_hat) / n) ** .5),
    else:
        print("The condition is not satisfied.")
        return
if method == "clopper_pearson":
    est = round(p_hat, 3)
    lwr = round(beta.ppf(alpha / 2, x, n - x + 1), 3)
    upr = round(beta.ppf(1 - alpha / 2, x + 1, n - x), 3)
if method == "Jeffrey":
    est = round(p_hat, 3)
    lwr = round(np.max([0, beta.ppf(alpha / 2, x + 0.5, n - x + 0.5)))
    upr = round(np.min([1, beta.ppf(1 - alpha / 2, x + 0.5, n - x
if method == "Agresti Coull":
    z = norm.ppf(1 - alpha / 2)
    n_{tilde} = n + z ** 2
    p_{tilde} = (x + z ** 2 / 2) / n_{tilde}
    est = round(p_tilde, 3)
    lwr = round(p\_tilde - z * ((p\_tilde * (1 - p\_tilde) / n) ** .5
    upr = round(p_tilde + z * ((p_tilde * (1 - p_tilde) / n) ** .5
if output == "string":
    return str(est) + '[' + str(level) + '%CI:(' + str(lwr) + ','
elif output == None:
    return {'est': est, 'lwr': lwr, 'upr': upr}
```

c.

```
In [18]: data = np.append(np.repeat(1, 42), np.repeat(0, 48))
```

confidence level	standard	normal approximation	clopper- pearson	Jeffrey's	Agresti-Coull
90%	0.467[90%CI:	0.467[90%CI:	0.467[90%CI:	0.467[90%CI:	0.468[90%CI:
	(0.38,0.553)]	(0.38,0.553)]	(0.376,0.559)]	(0.382,0.553)]	(0.381,0.554)]
95%	0.467[95%CI:	0.467[95%CI:	0.467[95%CI:	0.467[95%CI:	0.468[95%CI:
	(0.364,0.57)]	(0.364,0.57)]	(0.361,0.575)]	(0.366,0.569)]	(0.365,0.571)]
99%	0.467[99%CI:	0.467[99%CI:	0.467[99%CI:	0.467[99%CI:	0.469[99%CI:
	(0.331,0.602)]	(0.331,0.602)]	(0.331,0.606)]	(0.336,0.601)]	(0.333,0.604)]

The length of the different methods.

confidence level	standard	normal approximation	clopper-pearson	Jeffrey's	Agresti-Coull
90%	0.173	0.173	0.183	0.171	0.173
95%	0.206	0.206	0.214	0.203	0.206
99%	0.271	0.271	0.275	0.265	0.271

The Jeffrey's intervals have smallest width.