

# Report 3

```
import pandas as pd
import numpy as np
from distfit import distfit
import scipy.stats as ss
import matplotlib.pyplot as plt
```

## Chi-Square Function

```
""" data analysis with chi-square goodness of fit"""

def obs_cts(n, data):
    """ given: the data and number of bins
        returns: the observed values and the bin edges as lists"""
    events, edges = np.histogram(data, n)
    return events.tolist() , edges.tolist()

def exp_cts(n, data):
    """ given: the data and number of bins
        returns: the expected values and prob over each of the bins with
        the necessary modification of the first and last bins"""
    L=[]
    P_bins =[]
    for x in obs_cts(n,data)[1]:
        L.append(rv.cdf(x))
    P_bins.append(L[1])
    for i in range(1,len(L)-2):
        P_bins.append(L[i+1]-L[i])
    P_bins.append(1-L[-2])
    exp_cnt = [x * len(data) for x in P_bins]
```

```

    return exp_cnt, P_bins

def ind_bins_to_reduce(f_exp):
    """ given: a list
        returns: the indexes of the elements < 5 """
    NC_to_red =[index for index,value in enumerate(f_exp) if value < 5]
    return NC_to_red

def one_reduce(f_exp, f_obs, f_edge):
    """ given: lists of exp, obs, edges
        returns: new lists with one reduced bin with value < 5 """
    BTR = ind_bins_to_reduce(f_exp)

    if (len(BTR)>1 or (len(BTR)==1 and BTR[0]!=0)):
        f_exp[BTR[-1]-1] = f_exp[BTR[-1]-1]+f_exp[BTR[-1]]
        f_obs[BTR[-1]-1] = f_obs[BTR[-1]-1]+f_obs[BTR[-1]]
        del(f_edge[BTR[-1]])
        del(f_obs[BTR[-1]])
        del(f_exp[BTR[-1]])
    else:
        if BTR[0]==0:
            f_exp[1]= f_exp[1]+f_exp[0]
            f_obs[1]= f_obs[1]+f_obs[0]
            del(f_edge[1])
            del(f_obs[0])
            del(f_exp[0])

    f_expN = f_exp
    f_obsN = f_obs
    f_edgeN = f_edge
    BTRN = ind_bins_to_reduce(f_expN)
    return f_expN, f_obsN, f_edgeN, BTRN

def all_reduce(f_expF, f_obsF, f_edgeF, BTRF):
    """ finalizes the bin reduction """
    while BTRF !=[]:
        u = one_reduce(f_expF, f_obsF, f_edgeF)
        f_expF = u[0]
        f_obsF = u[1]
        f_edgeF = u[2]
        BTRF = u[3]

```

```

return f_expF, f_obsF, f_edgeF, BTRF

def model(data, n, dof):
    """ given data, the number of bins (n) and the number of estimated parameters (dof)
    produces the value of the chi-square test statistics and the p-value"""

    ## final expected count and final observed count after amalgamating bins
    exp, obs = all_reduce(exp_cts(n, data)[0], obs_cts(n, data)[0],
                          obs_cts(n, data)[1], ind_bins_to_reduce(exp_cts(n, data)[0]))[0:2]

    # build in chi-gof test, the last argument is the adjustment to the dof
    result = ss.chisquare( np.asarray(obs), np.asarray(exp), dof)
    return result

```

## Arrivals

### Graph Arrivals

```
data = pd.read_csv("Cafe Louis Data.csv")
```

```
data
```

	Observations	Inter-Arrival Time	Waiting Time	Service Time
0	1	0.00	0.00	51.90
1	2	15.38	43.96	35.75
2	3	68.47	0.00	25.13
3	4	13.09	0.00	33.16
4	5	19.40	11.14	31.19
...	...	...	...	...
246	247	3.85	9.45	38.54
247	248	0.85	44.04	65.76
248	249	10.97	45.04	12.31
249	250	0.59	71.38	102.32
250	251	61.44	66.43	49.42

```

arrivals = data["Inter-Arrival Time"]

dfit = distfit(distr = ['expon','erlang', 'gamma'])

dfit.fit_transform(arrivals)

[distfit] >INFO> fit
[distfit] >INFO> transform
[distfit] >INFO> [expon ] [0.00 sec] [RSS: 5.27859e-05] [loc=0.000 scale=27.675]
[distfit] >INFO> [erlang] [0.12 sec] [RSS: 0.000840505] [loc=-0.000 scale=672.021]
[distfit] >INFO> [gamma ] [0.10 sec] [RSS: 0.000110435] [loc=-0.000 scale=36.272]
[distfit] >INFO> Compute confidence intervals [parametric]

{'model': {'name': 'expon',
  'score': 5.278591824790488e-05,
  'loc': 0.0,
  'scale': 27.674940239043824,
  'arg': (),
  'params': (0.0, 27.674940239043824),
  'model': <scipy.stats._distn_infrastructure.rv_continuous_frozen at 0x2830a460390>,
  'bootstrap_score': 0,
  'bootstrap_pass': None,
  'color': '#e41a1c',
  'CII_min_alpha': 1.4195388568391432,
  'CII_max_alpha': 82.90671164278156},
'summary':
  name      score  loc      scale      arg \
0  expon  0.000053  0.0    27.67494      ()
1  gamma   0.00011 -0.0    36.272012 (0.6234259622109002,)
2  erlang   0.000841 -0.0   672.021451 (0.1599720453326794,)

                                params \
0                                (0.0, 27.674940239043824)
1 (0.6234259622109002, -5.8351220826759e-17, 36....
2 (0.1599720453326794, -2.866714624772024e-17, 6...

                                model bootstrap_score \
0 <scipy.stats._distn_infrastructure.rv_continuo...      0
1 <scipy.stats._distn_infrastructure.rv_continuo...      0
2 <scipy.stats._distn_infrastructure.rv_continuo...      0

bootstrap_pass  color

```

```

0          None #e41a1c
1          None #ff7f00
2          None #999999 ,
'histdata': (array([0.03327575, 0.02296495, 0.01921557, 0.01124814, 0.00843611,
0.00281204, 0.00234336, 0.00749876, 0.00187469, 0.00140602,
0.00093735, 0.00046867, 0.00046867, 0.00093735, 0.00140602,
0.00046867, 0.00046867, 0.          , 0.00046867, 0.          ,
0.          , 0.          , 0.          , 0.00046867, 0.          ,
0.          , 0.00046867])),
array([ 4.25037037, 12.75111111, 21.25185185, 29.75259259,
38.25333333, 46.75407407, 55.25481481, 63.75555556,
72.2562963 , 80.75703704, 89.25777778, 97.75851852,
106.25925926, 114.76          , 123.26074074, 131.76148148,
140.26222222, 148.76296296, 157.2637037 , 165.76444444,
174.26518519, 182.76592593, 191.26666667, 199.76740741,
208.26814815, 216.76888889, 225.26962963])),
'size': 251,
'alpha': 0.05,
'stats': 'RSS',
'bins': 'auto',
'bound': 'both',
'name': ['expon', 'erlang', 'gamma'],
'method': 'parametric',
'multtest': 'fdr_bh',
'n_perm': 10000,
'smooth': None,
'weighted': True,
'f': 1.5,
'n_boots': None,
'random_state': None}

```

```
dfit.plot()
```

```

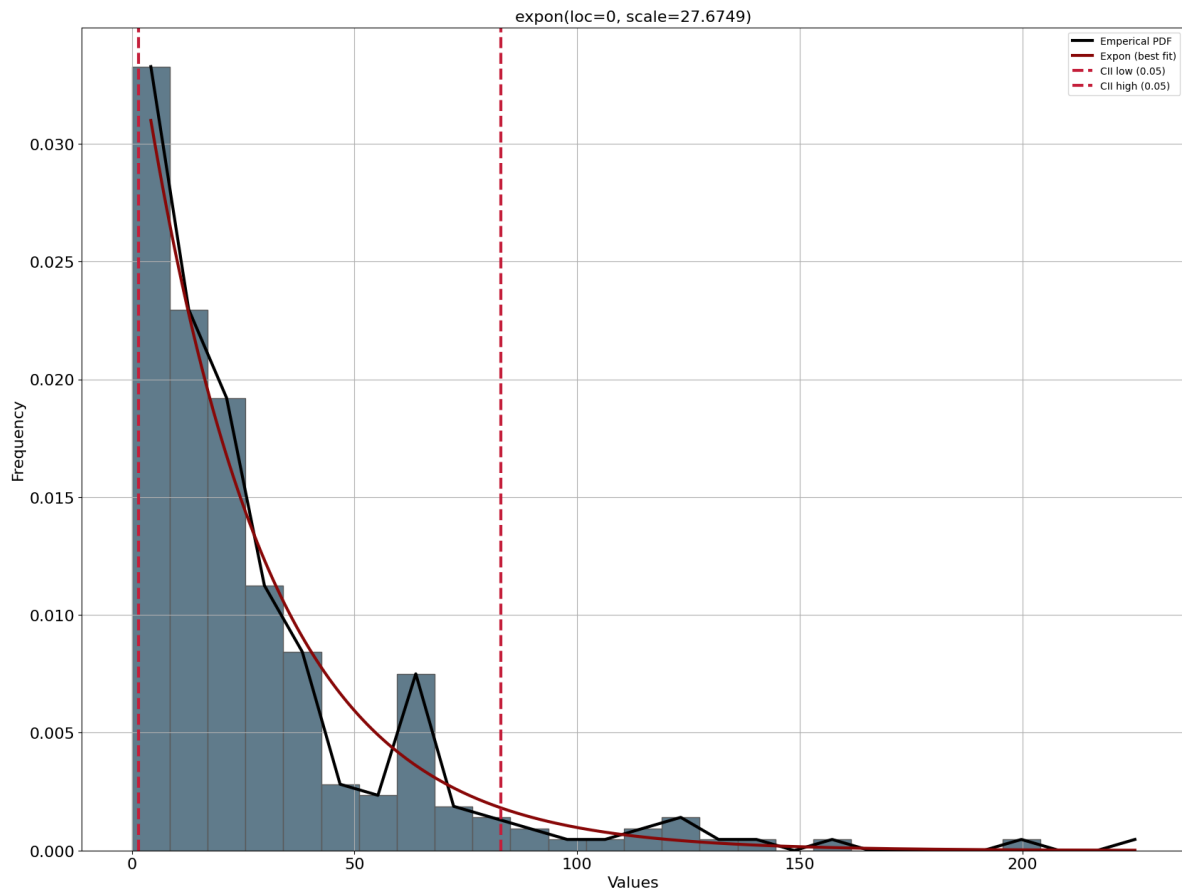
[distfit] >INFO> Create pdf plot for the parametric method.
[distfit] >INFO> Estimated distribution: Expon(loc:0.000000, scale:27.674940)

```

```

(<Figure size 2000x1500 with 1 Axes>,
 <Axes: title={'center': '\nexpon(loc=0, scale=27.6749)'}, xlabel='Values', ylabel='Frequency'>)

```



## Chi-Square Test

$H_0$  : The sample is exponentially distributed

$H_1$  : The sample is not exponentially distributed

```
# fitting data to exponential and estimating parameters

loc1, scale1 = ss.expon.fit(arrivals)

### Fit Exponential Distribution with estimated parameters.

rv = ss.expon(loc1,scale1)
```

```
### Chi-Square

model(arrivals,100,0)
```

```
Power_divergenceResult(statistic=64.83601758207016, pvalue=2.1836101023471424e-05)
```

## Service

### Graph

```
service = data["Service Time"]

dfit = distfit(distr = ['expon','erlang', 'gamma'])

dfit.fit_transform(service)
```

```
[distfit] >INFO> fit
[distfit] >INFO> transform
[distfit] >INFO> [expon ] [0.00 sec] [RSS: 0.000589297] [loc=5.970 scale=35.332]
[distfit] >INFO> [erlang] [0.10 sec] [RSS: 7.88516e-05] [loc=5.386 scale=16.879]
[distfit] >INFO> [gamma ] [0.10 sec] [RSS: 0.00273573] [loc=5.970 scale=6.358]
[distfit] >INFO> Compute confidence intervals [parametric]

{'model': {'name': 'erlang',
  'score': 7.88516486484426e-05,
  'loc': 5.385861580359169,
  'scale': 16.87921589450727,
  'arg': (2.127839859434259,),
  'params': (2.127839859434259, 5.385861580359169, 16.87921589450727),
  'model': <scipy.stats._distn_infrastructure.rv_continuous_frozen at 0x2830c52a590>,
  'bootstrap_score': 0,
  'bootstrap_pass': None,
  'color': '#e41a1c',
  'CII_min_alpha': 12.269739328734008,
  'CII_max_alpha': 88.93410586283824},
 'summary':
      name      score      loc      scale      arg \
0  erlang  0.000079  5.385862  16.879216  (2.127839859434259,)
1   expon  0.000589      5.97   35.332151      ()
2   gamma  0.002736      5.97   6.358043  (0.2983287282010668,)
```

```

                                params \
0 (2.127839859434259, 5.385861580359169, 16.8792...
1 (5.97, 35.33215139442231)
2 (0.2983287282010668, 5.969999999999999, 6.3580...

                                model bootstrap_score \
0 <scipy.stats._distn_infrastructure.rv_continuo... 0
1 <scipy.stats._distn_infrastructure.rv_continuo... 0
2 <scipy.stats._distn_infrastructure.rv_continuo... 0

bootstrap_pass    color
0      None    #e41a1c
1      None    #ff7f00
2      None    #999999 ,
'histdata': (array([0.0064361 , 0.01831812, 0.02574439, 0.02178371, 0.01633779,
                    0.00990169, 0.00495084, 0.00445576, 0.0064361 , 0.00297051,
                    0.00198034, 0.00148525, 0.00049508, 0.          , 0.00049508,
                    0.00148525, 0.          , 0.          , 0.          , 0.          ,
                    0.          , 0.00049508, 0.          , 0.          , 0.          ,
                    0.          , 0.          , 0.          , 0.00049508])),
array([ 9.99362069, 18.04086207, 26.08810345, 34.13534483,
        42.18258621, 50.22982759, 58.27706897, 66.32431034,
        74.37155172, 82.4187931 , 90.46603448, 98.51327586,
        106.56051724, 114.60775862, 122.655      , 130.70224138,
        138.74948276, 146.79672414, 154.84396552, 162.8912069 ,
        170.93844828, 178.98568966, 187.03293103, 195.08017241,
        203.12741379, 211.17465517, 219.22189655, 227.26913793,
        235.31637931])),
'size': 251,
'alpha': 0.05,
'stats': 'RSS',
'bins': 'auto',
'bound': 'both',
'name': ['expon', 'erlang', 'gamma'],
'method': 'parametric',
'multtest': 'fdr_bh',
'n_perm': 10000,
'smooth': None,
'weighted': True,
'f': 1.5,
'n_boots': None,
'random_state': None}

```



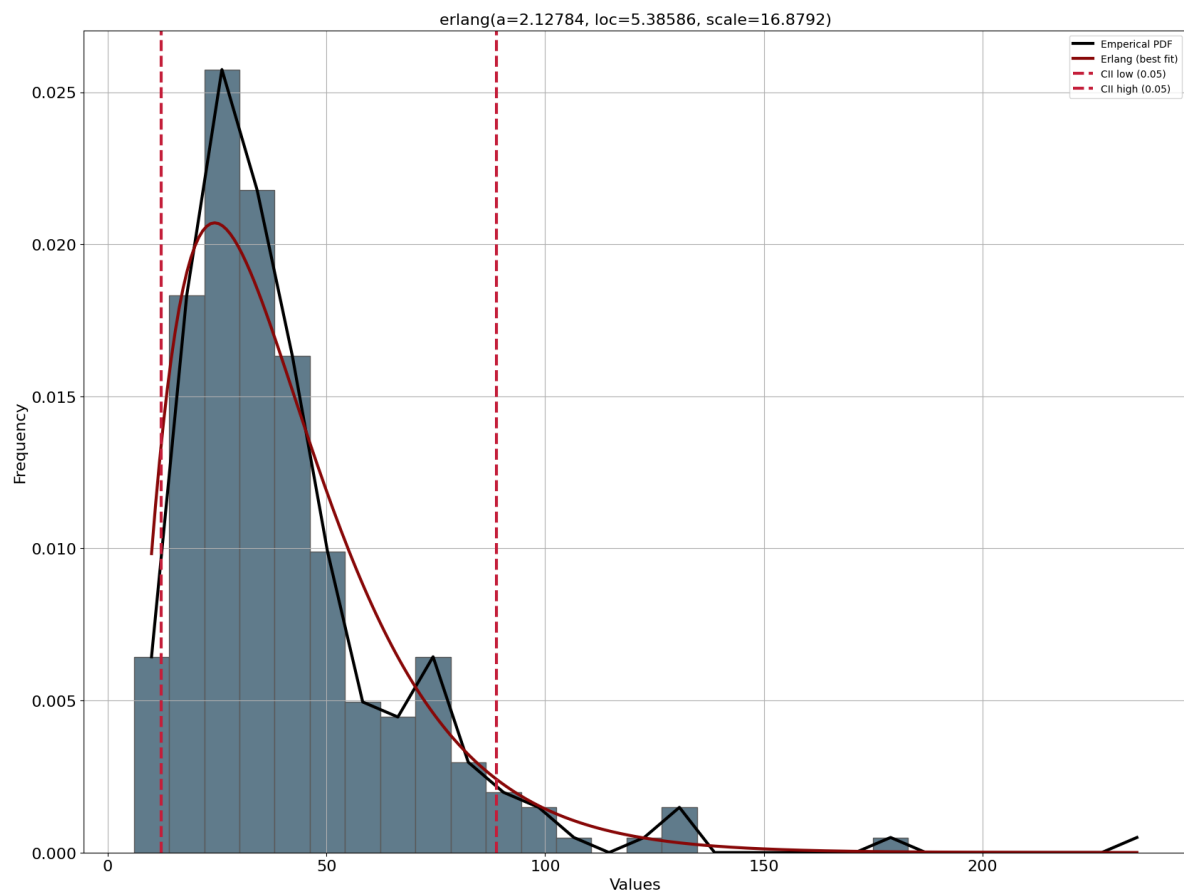
```
dfit.plot()
```

```
[distfit] >INFO> Create pdf plot for the parametric method.
```

```
[distfit] >INFO> Estimated distribution: Erlang(loc:5.385862, scale:16.879216)
```

```
(<Figure size 2000x1500 with 1 Axes>,
```

```
<Axes: title={'center': '\nerlang(a=2.12784, loc=5.38586, scale=16.8792)'}>, xlabel='Values'
```



## Chi-Square

- H0: The sample is erlang distributed
- H1: The sample is not erland distributed

```
### Parameter fit based on data

fit_k,fit_loc,fit_beta = ss.erlang.fit(service)

### Fit Erlang Distribution with estimated parameters.

rv = ss.erlang(fit_k,fit_loc,fit_beta)

### Chi-Square

model(service,100,0)
```

Power\_divergenceResult(statistic=26.492185948356493, pvalue=0.546009941220496)

## Result

P-value indicates we can accept the null hypothesis that the sample is erlang distributed.