```
In [1]:
```

```
# getting the libraries that has some of the functions needed for simulations
import numpy as np
import matplotlib.pyplot as plt
from numba import jit
import time
from numba import njit, prange
import pandas as pd
%matplotlib inline
```

# In [2]:

```
@jit(nopython=True)
def cdf inv fr(u, gamma):
 # function for getting inverse of CDF
 return ((pow(-np.log(u) , -gamma)))
@jit(nopython=True)
def cdf inv H(u , gamma, beta):
  #getting inverse of the function defined by H in the paper
  return (pow(-np.log(u)/(pow(beta , -1/gamma)-1) , -gamma))
@jit(nopython=True)
def armax(beta , gamma, n):
 # simulating the armax distribution
  # declaring an array
 x = np.zeros(n)
  # declaring uniform distribution
 u = np.random.uniform(0,1,1)[0]
  # applying cdf inv fr
 x0 = cdf inv fr(u,gamma)
 xi lag = x0
  x[0] = x0
  t = 1
  # getting the armax, ie, (max of Xi, Zi) * beta
  for i in range(n-1):
   u = np.random.uniform(0,1,1)[0]
    zi = cdf inv H(u, gamma, beta)
   xi = beta*max(xi_lag , zi)
   xi lag = xi
   x[t] = xi
    t = t + 1
  return x
```

#### In [3]:

```
@jit(nopython=True)
def theta_n_k(X, n, k_=1):
    sum = 0
    k = int(k_)

# getting the ascending n-k th order statistic
X_k = np.partition(X, n-k-1)[n-k-1]

for j in range(n-1):
    # k-th top order equals n-k low order
    if X[j] <= X_k and X[j+1] > X_k:
        sum += 1

if k == 0:
    return 1
    return sum/k
```

```
@jit(nopython=True)
def theta_GJ_k(X, n, k, delta):

numerator = (delta*delta + 1) * theta_n_k(X,n, int(np.floor(delta*k)) + 1) - delta*(th
eta_n_k(X, n, int(np.floor(delta*delta*k)) + 1) + theta_n_k(X, n, k))
denominator = (1 - delta)**2

if numerator < 0:
    return 0

return numerator/denominator</pre>
```

### In [4]:

```
@jit(nopython = True)
def np_apply_along_axis(funcld, axis, arr):
    # function as a workaround for a problem in calculating mean
    assert arr.ndim == 2
    assert axis in [0, 1]
    if axis == 0:
        result = np.empty(arr.shape[1])
        for i in range(len(result)):
            result[i] = funcld(arr[:, i])
    else:
        result = np.empty(arr.shape[0])
        for i in range(len(result)):
            result[i] = funcld(arr[i, :])
        return result
```

### In [5]:

```
@jit(nopython=True)
def simulate mean mse(n, theta, runs = 15, replicates = 10):
  all values mean = np.zeros((runs, n))
  all_values_mse = np.zeros((runs, n))
  for run in range(runs):
   a = np.zeros((replicates, n))
    for i in range(replicates):
      a[i] = armax(1-theta, 1, n)
   path = np.zeros((replicates, n))
   path2 = np.zeros((replicates, n))
    k range = list(range(n))
    for k in k range:
     for j in range(replicates):
        path[j][k] = theta_n_k(a[j], n, k)
        path2[j][k] = (path[j][k] - theta)**2
    all values mean[run] = np apply along axis(np.mean, 0, path)
    all_values_mse[run] = np_apply_along_axis(np.mean, 0, path2)
 return np apply along axis(np.mean, 0, all values mean), np apply along axis(np.mean,
0, all values mse)
```

# In [6]:

```
@jit(nopython=True)
def simulate_mean_mse_GJ(n, theta, delta = 0.25, runs = 15, replicates = 10):
    all_values_mean = np.zeros((runs, n))
    all_values_mse = np.zeros((runs, n))

for run in range(runs):
    a = np.zeros((replicates, n))
    for i in range(replicates):
        a[i] = armax(1-theta, 1, n)

    path = np.zeros((replicates, n))
```

```
path2 = np.zeros((replicates, n))

k_range = list(range(n))

for k in k_range:
    for j in range(replicates):
        path[j][k] = theta_GJ_k(a[j], n, k, delta)
        path2[j][k] = (path[j][k] - theta)**2

all_values_mean[run] = np_apply_along_axis(np.mean, 0, path)
    all_values_mse[run] = np_apply_along_axis(np.mean, 0, path2)

return np_apply_along_axis(np.mean, 0, all_values_mean), np_apply_along_axis(np.mean, 0, all_values_mse)
```

### In [7]:

```
def subsampling GJ(X, n, theta, T = 2, delta = 0.25):
 r = int(np.floor(n/T))
 theta_Vi = np.zeros((T, r))
  \#X = armax(1-theta, 1, n)
 for i in range(T):
   V = X[i:(r-1)*T+i+1:T]
   for j in range(1, r):
     theta_{Vi[i][j]} = theta_{GJ_k(V, V.shape[0], j, delta)}
  theta sub = -1 * np.ones(n)
  for j in range (1, r):
    \#temp = 1 - np.power(abs(1 - theta Vi[:,j]), 1/T).mean(axis = 0)
    theta sub[(j-1)*T+1] = 1 - ((np.power(1 - theta Vi[:,j], 1/T)).mean())
    theta_sub[np.isnan(theta_sub)] = 0
 for k in range(1, n):
   if theta_sub[k] == -1:
      theta sub[k] = theta sub[k-1]
  return theta sub
```

# In [8]:

```
@njit
def simulate_mean_mse_GJ_sub(n, T = 2, theta = 0.2, delta = 0.25, runs = 15, replicates = 10):
    all_values_mean = np.zeros((runs, n))
    all_values_mse = np.zeros((runs, n))

for run in range(runs):
    #print("\nrun = ", run + 1, "/", runs ,"...", sep = "", end = " ")
    a = np.zeros((replicates, n))
    for i in range(replicates):
        a[i] = armax(1-theta, 1, n)

path = np.zeros((replicates, n))
path2 = np.zeros((replicates, n))

k_range = list(range(n))

for j in range(replicates):
    path[j] = subsampling_GJ(a[j], n, theta, T, delta)
    path2[j] = np.power((path[j] - theta), 2)

#one_run_mean = np.mean(path, axis = 0)
```

```
all_values_mean[run] = np_apply_along_axis(np.mean, 0, path)
all_values_mse[run] = np_apply_along_axis(np.mean, 0, path2)

return np_apply_along_axis(np.mean, 0, all_values_mean), np_apply_along_axis(np.mean, 0, all_values_mse)
```

# In [9]:

```
@njit
def reff(mse_n, mse_gj):
    return np.sqrt(mse_n/mse_gj)

@njit
def bri(theta_n, theta_gj, theta):
    return np.abs((theta_n - theta)/(theta_gj - theta))

@njit
def sti(e_n, e_gj, theta, n, senstivity = 0.01):
    numerator = 0
    denominator = 0
    for i in range(n - 1):
        if np.abs(e_n[i] - theta) <= senstivity:
            denominator += 1
        if np.abs(e_gj[i] - theta) <= senstivity:
            numerator += 1
        return numerator/denominator</pre>
```

#### In [10]:

```
%%time
ns = [1000, 2000]
Ts = [2, 3, 4]
runs = 5000
replicates = 10
delta = 0.25
theta = 0.2
kns = np.zeros((len(Ts), len(ns)))
kgjs = np.zeros((len(Ts), len(ns)))
ens = np.zeros((len(Ts), len(ns)))
egjs = np.zeros((len(Ts), len(ns)))
mses = np.zeros((len(Ts), len(ns)))
msegjs = np.zeros((len(Ts), len(ns)))
reffs = np.zeros((len(Ts), len(ns)))
bris = np.zeros((len(Ts), len(ns)))
stis = np.zeros((len(Ts), len(ns)))
start = time.time()
i = 0
for T in Ts:
  j = 0
  for n in ns:
    print("T = \{0\} and n = \{1\}".format(T, n))
    e n, mse n = simulate mean mse(n, theta, runs, replicates)
    e_gj, mse_gj = simulate_mean_mse GJ sub(n, T, theta, delta, runs, replicates)
    k_n = np.where(mse_n == mse_n.min())[0][0]
    k gj = np.where(mse gj == mse gj.min())[0][0]
    print("k/n\t{0:.4f}".format((k n + 1)/n))
    print("k gj/n t{0:.4f}".format((k gj+1)/n))
    kns[i][j] = (k n + 1)/n
    kgjs[i][j] = (k gj + 1)/n
    print("E N \setminus t \{0:.4f\}".format(e n[k n]))
    print("E_GJ\t{0:.4f}".format(e_gj[k_gj]))
    ens[i][j] = e_n[k_n]
```

```
egjs[i][j] = e_gj[k_gj]
    print("MSE N\t{0:.4f}".format(mse n[k n]))
    print("MSE_GJ\t{0:.4f}".format(mse_gj[k_gj]))
    mses[i][j] = mse n[k n]
    msegjs[i][j] = mse\_gj[k\_gj]
    \label{eq:continuous_loss}  \texttt{print}(\texttt{"REFF} \setminus \texttt{t}\{0:.4f\}\texttt{".format}(\texttt{reff}(\texttt{mse}_n[\texttt{k}_n], \texttt{mse}_gj[\texttt{k}_gj]))) 
    \label{eq:print("BRI\t{0:.4f}".format(bri(e_n[k_n], e_gj[k_gj], theta)))} \\
    print("STI\t{0:.4f}".format(sti(e n, e gj, theta, n)))
    reffs[i][j] = reff(mse n[k n], mse gj[k gj])
    bris[i][j] = bri(e n[k n], e gj[k gj], theta)
    stis[i][j] = sti(e_n, e_gj, theta, n)
    end = time.time()
    print("Time =", end - start, "\n")
    j += 1
  print("\n\n")
  i += 1
T = 2 and n = 1000
k/n 0.1700
k gj/n 0.8720
E_N 0.1827
E GJ 0.2164
MSE_N 0.0011
MSE GJ 0.0020
REFF 0.7348
BRI 1.0539
STI 7.6667
Time = 1595.3145959377289
T = 2 and n = 2000
k/n 0.1355
k gj/n 0.8000
E N 0.1855
E GJ 0.2125
MSE N 0.0007
MSE GJ 0.0011
REFF 0.8089
BRI 1.1589
STI 8.6026
Time = 7409.737357378006
T = 3 and n = 1000
k/n 0.1680
k \, gj/n \, 0.8750
E N 0.1834
E GJ 0.2161
MSE N 0.0010
MSE GJ 0.0021
REFF 0.7031
BRI 1.0313
STI 6.8462
Time = 8807.554994344711
T = 3 and n = 2000
k/n 0.1380
k_{gj/n} 0.8155
E_N 0.1854
E GJ 0.2131
MSE N 0.0007
```

MSE\_GJ 0.0012 REFF 0.7729 BRI 1.1105 STI 7.9108

Time = 13815.05311369896

```
T = 4 and n = 1000
k/n 0.1710
k gj/n 0.8940
E_N 0.1827
E_GJ 0.2190
MSE_N 0.0011
MSE GJ 0.0025
REFF 0.6437
BRI 0.9122
STI 3.8462
Time = 15124.925018072128
T = 4 and n = 2000
k/n 0.1355
k_{gj/n} 0.8310
E N 0.1859
E GJ 0.2139
MSE N 0.0007
MSE GJ 0.0013
REFF 0.7263
BRI 1.0140
STI 7.8500
Time = 19753.08597445488
CPU times: user 5h 29min 13s, sys: 32.9 s, total: 5h 29min 46s
Wall time: 5h 29min 13s
In [11]:
np.save("kn1.npy", kns)
np.save("k_gj1.npy", kgjs)
np.save("e_n1.py", ens)
np.save("e_gj1.npy", egjs)
np.save("mse_n1.npy", mses)
np.save("mse gj1.npy", msegjs)
np.save("reff1.npy", reffs)
np.save("bri1.npy", bris)
np.save("stil.npy", stis)
In [ ]:
```