COMP 312 Assignment 6

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1 Python

1.1 Program A

1.1.1 Code

```
import random
import numpy
import math
## Useful extras ----
\mathbf{def} \ \operatorname{conf}(\mathbf{x}):
    """95%_confidence_interval"""
    lower = numpy.mean(x) - 1.96*numpy.std(x)/math.sqrt(len(x))
    upper = numpy.mean(x) + 1.96*numpy.std(x)/math.sqrt(len(x))
    return (lower, upper)
def coxian():
         """ Generates _a _coxian _random _ variate """
         t = random.expovariate(2)
         if random () \leq 0.2:
                  return t
         t += random.expovariate(3)
         return t
if _-name_- = "_-main_-":
         random.seed(123)
         a = []
         b = []
         for i in range (10000):
                  variate = coxian()
                  a.append(variate)
                  b.append(math.pow(variate, 2))
         print "E[t]:", conf(a)
         print "E[t^2]:", conf(b)
1.1.2 Output
E[t]: (0.75741101633238683, 0.78099765992129067)
E[t^2]: (0.92233300526267215, 0.98510362267500762)
The theoretical values are within the 95\% confidence interval which means we
```

can conclude that E[t] and $E[t^2]$ are equal to the theoretical values

1.2 Program B

1.2.1 Code

```
""" (q6.py) \( \text{M/G/c_queueing_system_with_service_time_monitors} \) ""
from SimPy.Simulation import *
import random
import numpy
import math
import coxian
## Useful extras ----
\mathbf{def} \ \operatorname{conf}(\mathbf{x}):
    """95%_confidence_interval"""
    lower = numpy.mean(x) - 1.96*numpy.std(x)/math.sqrt(len(x))
    upper = numpy.mean(x) + 1.96*numpy.std(x)/math.sqrt(len(x))
    return (lower, upper)
## Model -
class Source(Process):
    """ generate_random_arrivals"""
    def run(self, N, lamb):
         for i in range (N):
             a = Arrival(str(i))
             activate(a, a.run())
             t = random.expovariate(lamb)
             yield hold, self, t
class Arrival(Process):
    """ an _ arrival"""
    n = 0
    def run(self):
         Arrival.n += 1
        G. numbermon. observe (Arrival.n)
         arrivetime = now()
         yield request, self, G. server
         t = coxian.coxian()
        G. servicemon. observe(t)
        G. servicesquaredmon.observe(t**2)
         yield hold, self, t
         yield release, self, G. server
         delay = now() - arrivetime
```

```
G. delaymon. observe (delay)
        #print now(), "Observed delay", delay
        Arrival.n -= 1
        G. numbermon. observe (Arrival.n)
class G:
    server = 'dummy'
    delaymon = 'Monitor'
    numbermon = 'Monitor'
    servicemon = 'Monitor'
    servicesquaredmon = 'Monitor'
def model(c, N, lamb, maxtime, rvseed):
    \# setup
    initialize()
    random.seed(rvseed)
    G. server = Resource(c, monitored=True)
    G. delaymon = Monitor()
    G.numbermon = Monitor()
    G. servicemon = Monitor()
    G.servicesquaredmon = Monitor()
    \# simulate
    s = Source('Source')
    activate(s, s.run(N, lamb))
    simulate (until=maxtime)
    # gather performance measures
    L = G.numbermon.timeAverage()
   LQ = G. server.waitMon.timeAverage()
   W = G. delaymon.mean()
    S = G. servicemon.mean()
    S2 = G. services quared mon. mean()
    lambEff = L/W
   WQ = LQ/lambEff
    row = lambEff * S
    return (WQ, lambEff, S2, row)
## Experiment ----
allY = []
for i in range (50):
```

We want Y to be 0 as this would show $w_q = \frac{\lambda E[t^2]}{2(1-\rho)}$. We see that 0 is within the confidence interval of the average Y. Not only this but our mean Y is small enough for us to conclude that $w_q = \frac{\lambda E[t^2]}{2(1-\rho)}$

Conf Y: (-0.011739640704165668, 0.0070782512379870006)