```
In [34]:
```

```
import numpy as np
import scipy as sp
from scipy import optimize
import matplotlib.pyplot as plt
from sympy.solvers import solve
from sympy import Symbol
import math as mt
```

Problem 1

```
In [5]:
```

```
#define parameters
R = 5.0
C = 2.0
mu = 10.0
lambd = 8.0
rho = lambd/mu
```

In [17]:

```
#define social rate function
def f(n):
    return lambd*R*((1-rho**n)/(1-rho**(n+1)))-(C*((rho/(1-rho))-((n+1)*rho**(n+1)))/(1-rho**(n+1))))
```

```
In [29]:
#plot expected social benefit rate vs n
for n in range(1,40,1):
    plt.plot(n, f(n), marker = "o")
plt.title('expected social benefit rate vs n')
plt.show()
           expected social benefit rate vs n
          .....
 32
 30
 28
 26
 24
 22
                       20
                            25
             10
                  15
                                 30
                                      35
In [19]:
#find the exact max value and associated n value, define the benefit
#rate again
def SO(n):
    return lambd*R*((1-rho**n[0])/(1-rho**(n[0]+1)))-C*((rho/(1-rho))-((n[0]+1)*
rho**(n[0]+1))/(1-rho**(n[0]+1)))
In [20]:
sp.optimize.fmin(lambda n: -SO(n), np.array([0,0]))
Optimization terminated successfully.
         Current function value: -33.240874
         Iterations: 56
         Function evaluations: 96
```

As shown above result, when $n^* \approx 8$, the social benefit rate reaches maximum where the social maximum rate is about 33.24.

Out[20]:

array([7.86098027, -3.93528052])

```
In [12]:
#calculate individual optimize system size
n_e = (R * mu)/C
print(n_e)
```

25.0

 n^* = 10 < n_e = 25, socially optimal system size is less than the individually optimal system size n_e .

```
In [22]:

p = Symbol('p')
solve(((R - p)*mu)/C - 8, p)

Out[22]:
```

[3.40000000000000]

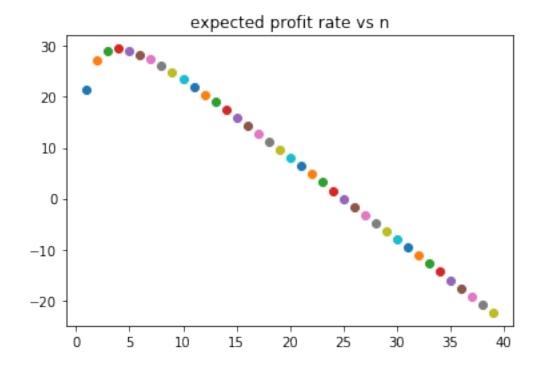
when p = 3.4, $n^* = n_e$, the lowest toll is 3.4.

Problem 2

```
In [23]:
```

```
def z_n(n):
    return lambd*((1-rho**n)/(1-rho**(n+1)))*(R - (C*n)/mu)
```

```
for n in range(1,40,1):
    plt.plot(n, z_n(n), marker = "o")
plt.title('expected profit rate vs n')
plt.show()
```



```
In [27]:
```

```
#define the profit function again in array form for optimizer
def z_n_a(n):
    return lambd*((1-rho**n[0])/(1-rho**(n[0]+1)))*(R - (C*n[0])/mu)
```

```
In [28]:
```

```
sp.optimize.fmin(lambda n: -z_n_a(n), np.array([0,0]))
```

```
Optimization terminated successfully.
```

Current function value: -29.507398

Iterations: 49

Function evaluations: 84

Out[28]:

array([3.93776678, -1.97132956])

 $n_m \approx$ 4, $n_m < n^*$

Problem 4

```
In [145]:
    rho4 = 0.95
    def p0_cal(c):
        sec_part = 0
        for j in range(0, c, 1):
            sec_part = sec_part + ((c*rho4)**j)/mt.factorial(j)
        return (((c*rho4)**c)/((1-rho4)*mt.factorial(c)) + sec_part)**(-1)

In [151]:

for c in [1, 2, 4, 8, 16, 32]:
        p0 = p0_cal(c)
        d_frac = p0*((c**c)/mt.factorial(c))*((rho4**c)/(1 - rho4))
        d_exp = (1/(1 - rho4*(1/c))) * d_frac
        print('c = ', c, ",", 'fraction delayed = ', d_frac, ', expected delay time
        = ', d_exp)

c = 1 , fraction delayed = 0.95 , expected delay time = 18.999999
```

```
c = 1 , fraction delayed = 0.95 , expected delay time = 18.999999
999999992
c = 2 , fraction delayed = 0.9256410256410256 , expected delay tim
e = 1.7631257631257629
c = 4 , fraction delayed = 0.8914189951776734 , expected delay tim
e = 1.1690740920362932
c = 8 , fraction delayed = 0.844168630609684 , expected delay time
= 0.9579218503372301
c = 16 , fraction delayed = 0.780192462248371 , expected delay tim
e = 0.8294404914268395
c = 32 , fraction delayed = 0.6957101141665399 , expected delay ti
me = 0.7169959308640669
```

In []: