```
In [1]: import numpy as np
import matplotlib.pyplot as plt
```

Problem 3.1

Part A

182.68823748261997

Part B

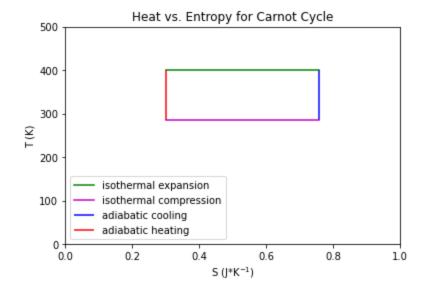
```
In [3]: #work done by engine
w_carnot_cycle = -n*R*Th*np.log(Va/Vb) - n*R*Tc*np.log(Vc/Vd) #Work in Joules
print(w_carnot_cycle)
```

52.98753986550585

Part D

```
In [4]: S_a = 0.3 \#J/K
        DeltS_Th = n*R*np.log(Vb/Va) #J/K
        S_b = DeltS_Th + S_a \#J/K
        DeltS_Tc = n*R*np.log(Vd/Vc) #J/K
        S c = S b
        S_d = DeltS_Tc + S_c
        Sh = np.linspace(S_a, S_b, 100)
        Ths = [Th for x in range(100)]
        Sc = np.linspace(S_c, S_d, 100)
        Tcs = [Tc for x in range(100)]
        plt.plot(Sh, Ths, label="isothermal expansion", color = 'g')
        plt.vlines(x = S_b, ymax = Th, ymin = Tc, color = 'b', label="adiabatic cooling")
        plt.plot(Sc, Tcs, label="isothermal compression", color = 'm')
        plt.vlines(x = S_d, ymax = Th, ymin = Tc, color = 'r', label="adiabatic heating")
        plt.axis([0,1,0,500])
        plt.ylabel("T (K)")
        plt.xlabel("S (J*K$^{-1}$)")
        plt.legend()
        plt.title("Heat vs. Entropy for Carnot Cycle")
```

Out[4]: Text(0.5, 1.0, 'Heat vs. Entropy for Carnot Cycle')



Part E

```
In [5]: A = (S_b - S_a) * (Th - Tc) #Simple area formula: Delta_Y * Delta_X
print(A) #J (about equal to work!)
```

52.75122857310652

Problem 3.2

```
In [6]: T_initial = 25 + 273.15 #C -> K
T_final = 100 + 273.15 #C -> K
V_initial = 0.5 * 0.001 #L -> m^3
V_final = 1 * 0.001 #L -> m^3
P_initial = 1 * 101325 #Atm -> Pa
n = P_initial * V_initial / (R * T_initial)
Cv = (3/2) * n * R #Assume Ar behaves as a monoatomic idg
Cp = Cv + n*R

Del_S_isochoric = Cv*np.log(T_final / T_initial)
Del_S_isobaric = Cp*np.log(V_final / V_initial)
Del_S_tot = Del_S_isochoric + Del_S_isobaric
print(Del_S_tot) #Total Entropy Change in J/K
```

0.3516457707793893

Problem 3.3

Part A

```
In [7]: Ntot = 10
        Vtot = 60
        nL = 10
        VL = 10
        nR = 20
        VR = Vtot - VL
        def n_choose_k(n, k):
            return np.math.factorial(n) / (np.math.factorial(k) * np.math.factorial(n-k))
        def Boltzmann_Entropy(W):
            Avog = 6.022e23
            return (R/Avog)*np.log(W)
        def Find_Entropy_Stats(nL, ntot, VL, Vtot):
            W_Left = n_choose_k(VL, nL) #Multiplicity of Left
            W_Right = n_choose_k(Vtot-VL, ntot-nL) #Multiplicity of Right
            W_Tot = W_Left * W_Right
            S_Left = Boltzmann_Entropy(W_Left)
            S_Right = Boltzmann_Entropy(W_Right)
            S_Total = S_Left + S_Right
            return (W_Left, W_Right, W_Tot, S_Left, S_Right, S_Total)
        print("Left Mult.:\t{}\nRight Mult.:\t{}\nTotal Mult.:\t{}\n\nLeft S:\t\t{}\nRight S:\t
        {}\nTotal S:\t{}".format(*Find_Entropy_Stats(nL, nL+nR, VL, Vtot)))
```

Left Mult.: 1.0

Right Mult.: 47129212243960.0 Total Mult.: 47129212243960.0

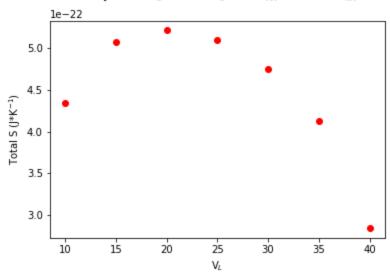
Left S: 0.0

Right S: 4.346944605367547e-22 Total S: 4.346944605367547e-22 The equilibrium position will lie where entropy (S) is maximized. From the plot shown below, this is at V_L =20 and V_R =40

```
In [8]: Vls = [x*5 + 10 for x in range(0,7)]
    S_total_vals = [Find_Entropy_Stats(nL, nL+nR, y, Vtot)[5] for y in Vls]
    plt.plot(Vls, S_total_vals, 'or')
    plt.xlabel("V$_L$")
    plt.ylabel("Total S (J*K$^{-1}$)")
    plt.title("Entropy Dependence on Number of Cells in \nLeft Subsystem (V$_L$) where n$_L$
    $=10, n$_{tot}$=30, and V$_{tot}$=60 \n")
```

Out[8]: Text(0.5, 1.0, 'Entropy Dependence on Number of Cells in \nLeft Subsystem (V L) where n L =10, n L =10, n L =30, and V L =60 \n')

Entropy Dependence on Number of Cells in Left Subsystem (V_L) where n_L =10, n_{tot} =30, and V_{tot} =60



Problem 3.5

Part C

Interpretation:

The entropy of the system increases up until around E_{tot} =6 and then begins to decrease. The corresponding Temperatures for a change in E (always 1) and the change in entropy calculated below is maximized when Entropy is maximized which is reflected in the dS being smallest around E_{tot} =6. Since internal energy changes are constantly 1 on the Y axis, for this system, large changes in S correspond to low temperatures and smaller changes in S correspond to high temperatures. This is consistent with what we should expect from the given relationship T = dU/dS. Small entropy changes reflect high temperature when internal energy is low, but when it is high, larger entropy changes are required to achieve the same temperature.

```
E_{\text{tots}} = [3, 4, 5, 6, 7, 8, 9]
In [9]:
         W_{\text{tots}} = [1, 3, 6, 7, 6, 3, 1]
         T_{tots} = []
         Boltzmann_S = [R/6.022e23 * np.log(x) for x in W_tots]
         prev = 0
         dS = []
         for i in Boltzmann_S:
              sub = i - prev
              dS.append(sub)
              prev = i
         T = [abs(1/j) \text{ for } j \text{ in } dS \text{ if } j != 0]
         plt.plot(Boltzmann_S, E_tots, 'r-o')
         plt.ylabel("E$_{tot}$")
         plt.xlabel("S (J*K$^{-1}$)")
         plt.title("Plot of E$_{tot}$ [U] vs. Entropy for System of 3 Molecules and Energy Level
         s")
         print("Temperature From Changes:")
         print(T)
```

Temperature From Changes:

[6.592652141135143e+22, 1.0449106423998241e+23, 4.698499328891361e+23, 4.698499328891361e+23, 1.0449106423998241e+23, 6.592652141135143e+22]

