

1.

Decision variables for our problem is the following:

i: Represents the employee number assignment

j: Represents the task number assignment

 x_{ij} for $i = 1, \dots, n$ $j = 1, \dots, n$ c_{ij} for $i = 1, \dots, n$ $j = 1, \dots, n$

$$\min \sum_{i=1}^n \sum_{j=1}^n c_{ij} x_{ij}$$

$$\text{s.t. } \sum_{i=1}^n x_i = 1 \text{ (Meaning each row sums to 1 for employee to task assignment)}$$

$$\text{s.t. } \sum_{j=1}^n x_j = 1 \text{ (Meaning each column sums to 1 for task to employee assignment)}$$

$$\text{s.t. } \sum_{i=1}^n \sum_{j=1}^n x_{ij} \in [0, 1] \text{ Boolean observations for each } x_{ij}$$

```
In [132]: import numpy as np
import cvxpy as cp

c = np.matrix([[3, 0, 0, 0], [7, 4, 4, 0], [11, 4, 10, 6], [8, 6, 9, 5]])
x = cp.Variable(c.shape, boolean=True)
constraints = [x>=0, x<=1]
objective = cp.Minimize(cp.sum(cp.sum(cp.multiply(c,x))))
prob = cp.Problem(objective, constraints)

prob.solve()
print(prob.value)
print(prob.status)

77.0
optimal

c:\users\mjpearl\desktop\omsa\isye-6669-oan\env_isye6669\lib\site-packages\ipykernel_launcher.py:4: PendingDeprecationWarning:
the matrix subclass is not the recommended way to represent matrices or deal with linear algebra (see https://docs.scipy.org/doc/numpy/user/numpy-for-matlab-users.html). Please adjust your code to use regular ndarray.
after removing the cwd from sys.path.
```

Figure 1: CVXPY Output

2.

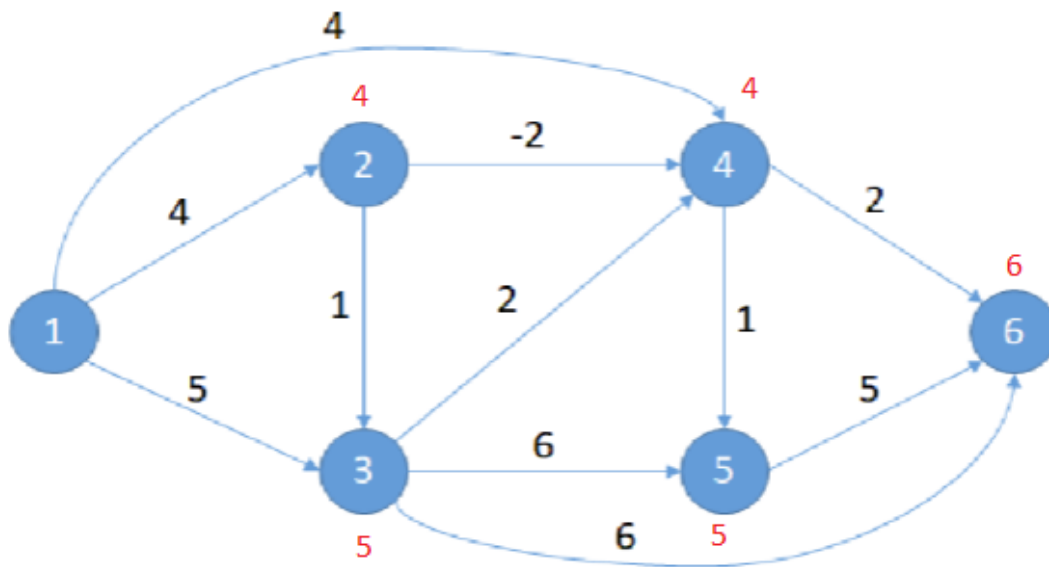


Figure 2: Shortest Path Bellman's Output

In our case the Bellman's output is in red for each corresponding node.

3.
a)

$$\min c_1 p_1 + c_2 p_2 + c_3 p_3 = 10p_1 + 6p_2 + 5c_3$$

s. t

$$-f_{61} + f_{12} = p_1$$

$$-f_{23} + f_{34} = p_2$$

$$-f_{45} + f_{56} = p_3$$

$$f_{12} - f_{23} = d_1 = 10$$

$$f_{34} - f_{45} = d_2 = 120$$

$$f_{56} - f_{61} = d_3 = 90$$

$$f_{12} = B_{12}(\theta_1 - \theta_2) = 11.6(\theta_1 - \theta_2)$$

$$f_{23} = B_{23}(\theta_2 - \theta_3) = 5.9(\theta_2 - \theta_3)$$

$$f_{34} = B_{34}(\theta_3 - \theta_4) = 13.7(\theta_3 - \theta_4)$$

$$f_{45} = B_{45}(\theta_4 - \theta_5) = 9.8(\theta_4 - \theta_5)$$

$$f_{56} = B_{56}(\theta_5 - \theta_6) = 5.6(\theta_5 - \theta_6)$$

$$f_{61} = B_{61}(\theta_6 - \theta_1) = 10.5(\theta_6 - \theta_1)$$

$$-f_{12}^{max} = -100 \leq f_{12} \leq f_{12}^{max} = 100$$

$$-f_{23}^{max} = -120 \leq f_{23} \leq f_{23}^{max} = 120$$

$$-f_{34}^{max} = -50 \leq f_{34} \leq f_{34}^{max} = 50$$

$$-f_{45}^{max} = -90 \leq f_{45} \leq f_{45}^{max} = 90$$

$$-f_{56}^{max} = -60 \leq f_{56} \leq f_{56}^{max} = 60$$

$$-f_{61}^{max} = -50 \leq f_{61} \leq f_{61}^{max} = 50$$

$$p_1^{min} = 20 \leq p_1 \leq p_1^{max} = 70$$

$$p_2^{min} = 20 \leq p_2 \leq p_2^{max} = 150$$

$$p_3^{min} = 10 \leq p_3 \leq p_3^{max} = 150$$

b-c)

Optimal value is in first part of the image, last cell contains the values at Nodes 2, 4 and 6.

Optimal Value: 1249.9999997631494
With the following variables: 19.999999928688347 50.00000011972041 149.99999995158868 59.85392656833289 63.21238250844543 71.5103636788361 71.43431176537509 59.7820689114058 54.238870488904865 -38.95808890530549 -48.958088905305054 1.041911214415802 -119.95808878558378 31.041911166005285 -58.95808883399427

```
In [129]: print(constraints[15].dual_value)
          print(constraints[16].dual_value)
          print(constraints[17].dual_value)
```

```
-5.9999999735722565
-5.999999975933699
-5.999999973803812
```