

```
In [2]: import numpy as np
import cvxpy as cp
import matplotlib.pyplot as plt
```

A4.25) Probability Bounds

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We can create a 4-D tensor with the axes representing X_1, X_2, X_3, X_4 . Then, each entry (x_1, x_2, x_3, x_4) is equal to the joint probability of $X_1 = x_1, X_2 = x_2, X_3 = x_3, X_4 = x_4$.

```
In [ ]: z = np.zeros((2,2,2,2))
z.sum(axis=(0, 3))[0, 1, 0]
```

```
Out[ ]: np.float64(0.0)
```

```
In [22]: z = np.ones((3, 1))
z.sum(axis=0)
```

```
Out[22]: array([3.])
```

```
In [34]: X = cp.Variable((2, 2, 2, 2))
constr425 = []

constr425 += [
    X >= 0,
    cp.sum(X) == 1,
    cp.sum(X[1, :, :, :]) == 0.9,
    cp.sum(X[:, 1, :, :]) == 0.9,
    cp.sum(X[:, :, 1, :]) == 0.1
]

#  $P(X_1 = 1, X_4 = 0 \mid X_3 = 1) = 0.7$ 
constr425 += [
    cp.sum(X[1, :, 1, 0]) == 0.7 * cp.sum(X[:, :, 1, :])
]

#  $P(X_4 = 1 \mid X_2 = 1, X_3 = 0) = 0.6$ 
constr425 += [
    cp.sum(X[:, 1, 0, 1]) == 0.6 * cp.sum(X[:, 1, 0, :])
]

prob425min = cp.Problem(cp.Minimize(cp.sum(X[:, :, :, 1])), constr425)
prob425min.solve()
print("Minimum possible value of  $P(X_4=1)$ : ", prob425min.value)
```

Minimum possible value of $P(X_4=1)$: 0.47999999984476227

C:\Users\kyler\AppData\Local\Temp\ipykernel_22044\1265068490.py:23: UserWarning: The problem has an expression with dimension greater than 2. Defaulting to the SCIPY backend for canonicalization.
prob425min.solve()

```
In [35]: prob425max = cp.Problem(cp.Maximize(cp.sum(X[:, :, :, 1])), constr425)
prob425max.solve()
print("Maximum possible value of P(X4=1): ", prob425max.value)
```

Maximum possible value of P(X4=1): 0.6100000000407165

C:\Users\kyler\AppData\Local\Temp\ipykernel_22044\2155099936.py:2: UserWarning: The problem has an expression with dimension greater than 2. Defaulting to the SCIPY backend for canonicalization.

```
prob425max.solve()
```

A6.13) Fitting Censored Data

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a) We can find c and $y^{M+1} \dots y^K$ that minimizes J by creating a constraint that ensures that $c^T x_i \geq D$ for $i \in [M, K]$, since we know the true value of the censored values is always greater than the upper bound D .

```
In [3]: np.random.seed(15)

n = 20; # dimension of x's
M = 25; # number of non-censored data points
K = 100; # total number of points
c_true = np.random.randn(n,1)
X = np.random.randn(n,K)
y = np.dot(np.transpose(X),c_true) + 0.1*(np.sqrt(n))*np.random.randn(K,1)

# Reorder measurements, then censor
sort_ind = np.argsort(y.T)
y = np.sort(y.T)
y = y.T
X = X[:, sort_ind.T]
D = (y[M-1]+y[M])/2.0
y = y[list(range(M))]
```

```
In [18]: X.shape, y.shape
```

```
Out[18]: ((20, 100, 1), (25, 1))
```

```
In [37]: c = cp.Variable((n, 1))
u = cp.Variable(K-M) # add unknown value to upper bound
obj613 = cp.Minimize(cp.sum_squares(y - X[:, :M].T @ c))
constr613 = [
    X[:, M:].T @ c >= D
]
prob613 = cp.Problem(obj613, constr613)
prob613.solve()
c_hat = c.value
print("c_hat = ", c.value.T)
```

```
C:\Users\kyler\AppData\Local\Temp\ipykernel_29640\877152390.py:8: UserWarning: The problem includes expressions that don't support CPP backend. Defaulting to the SCIPY backend for canonicalization.
  prob613.solve()
```

```
c_hat_ls = [[-0.86949623  0.38779257 -0.07924954 -0.52689695  0.44848029 -2.1460358
             -0.79172669 -0.86627365 -0.18321402 -0.25059169 -0.15833986  0.55531191
              0.42815852  0.06903265 -0.43355176  0.35683189 -0.20242313  2.00710944
              0.81069272  0.09358954]]
```

```
In [39]: print("c_hat error: ", np.linalg.norm(c_true - c_hat, 2) / np.linalg.norm(c_true, 2))
```

c_hat error: 0.1611548300994136

```
In [40]: print("c_hat_ls error: ", np.linalg.norm(c_true - c_hat_ls, 2) / np.linalg.norm(c_true, 2))
```

c_hat_ls error: 0.3332218561062694

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```
In [41]: # data for problem on bandlimited signal recovery from zero crossings
import numpy as np
n = 2048
f_min = 4
B = 9
s = np.array([-1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1, -1,
y = np.array([-1.0094, -0.9767, -0.9433, -0.9090, -0.8740, -0.8383, -0.8019, -0.764
```

A19.10) Scheduling

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a) We can make the objective a convex combination of C and T- for example, minimize $\theta C + (1 - \theta)T$ s.t. $\theta \in [0, 1]$, and then solve for various values of θ .

```

In [43]: import numpy as np
import matplotlib.pyplot as plt

np.random.seed(1)

# total number of tasks
n = 10

alpha = np.random.rand(n) + 1
m = np.random.rand(n) + 1

P = [
    [0, 6], [1, 2], [1, 3], [1, 4], [1, 6],
    [2, 7], [3, 7], [3, 8], [5, 9], [6, 9], [8, 9]
]

def visualize_schedule(s, f, save=None):
    """
    Visualize a schedule.

    Parameters
    -----
    s : numpy.ndarray
        Start times of tasks.
    f : numpy.ndarray
        Finish times of tasks.
    save : str, optional
        Path to save the figure.
    """

    fig, ax = plt.subplots(figsize=(8, 4))
    ax.set_position([0.1, 0.1, 0.7, 0.8])

    # compute cost of each task
    cost = [alpha[k] / (f[k] - s[k] - m[k]) for k in range(n)]
    colorbar_max_cost = np.round(1.5 * max(cost), 1)

    # order of display of tasks
    order = np.array([0, 6, 1, 4, 5, 2, 7, 3, 8, 9])

    # plot tasks as rectangles
    for i, k in enumerate(reversed(order)):
        color = plt.cm.BuPu(cost[k] / colorbar_max_cost)
        ax.fill([s[k], f[k], f[k], s[k]], [i, i, i + 1, i + 1], color=color, edgecolor='black')
        ax.text((s[k] + f[k]) / 2, i + 0.5, str(k + 1), fontsize=12, ha='center', va='bottom')

    # plot precedence constraints
    for i, j in P:
        for k in [i, j]:
            ypos = n - np.argwhere(order == k)[0, 0]
            ax.plot([f[i], f[j]], [ypos - 1, ypos], color='red', linewidth=1.5)

    # axes
    T = max(f)

```

```

ax.set_xlim([0, T])
ax.set_xticks(np.arange(0, T + 1, 5))
ax.set_xlabel('time')
ax.set_ylim([0, n])
ax.set_yticks([])
ax.set_ylabel('task')

# colorbar
ax_cbar = fig.add_axes([0.85, 0.1, 0.02, 0.8])
ax_cbar.imshow(np.linspace(0, 1, 256)[: , None], aspect='auto', origin='lower',
ax_cbar.set_xticks([])
ax_cbar.yaxis.tick_right()
ax_cbar.set_yticks([0, 256])
ax_cbar.set_yticklabels([0, colorbar_max_cost])
ax_cbar.set_title('cost', fontsize=10)

# render
if save is not None:
    plt.savefig(save, bbox_inches='tight')
plt.show()

```

```

In [63]: eps = 1e-6
s = cp.Variable(n)
f = cp.Variable(n)
T = cp.Parameter()

ts = []
costs = []

for T_val in range(10, 31):

    # Objective: only the convex cost
    objective = cp.Minimize(
        cp.sum(cp.multiply(alpha, cp.inv_pos(f - s - m)))
    )

    constraints = [
        f - s >= m + eps,
        s >= 0,
        f >= 0,
        cp.max(f) <= T_val
    ]

    # precedence constraints
    for (i, j) in P:
        constraints += [s[j] >= f[i]]

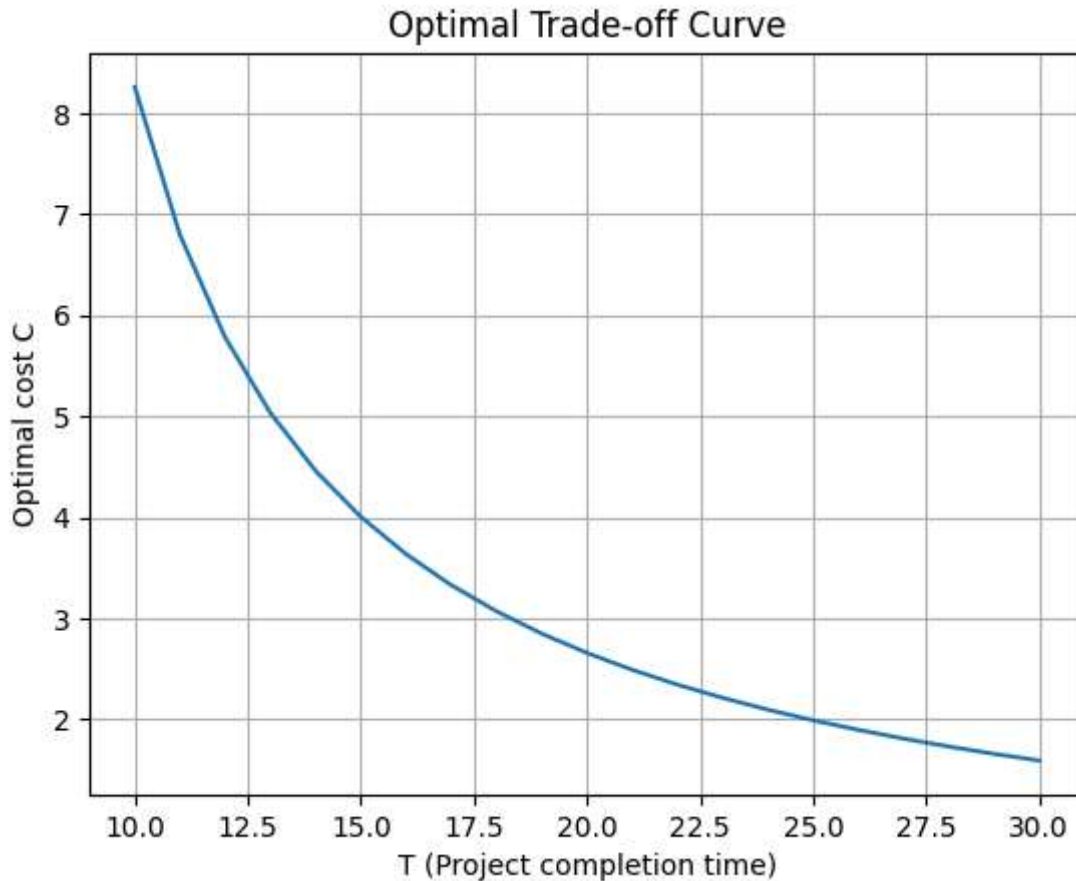
    prob = cp.Problem(objective, constraints)
    prob.solve()

    ts.append(T_val)
    costs.append(prob.value)

# Plot
plt.plot(ts, costs)

```

```
plt.xlabel("T (Project completion time)")
plt.ylabel("Optimal cost C")
plt.title("Optimal Trade-off Curve")
plt.grid(True)
plt.show()
```



```
In [62]: objective = cp.Minimize(
    cp.sum(cp.multiply(alpha, cp.inv_pos(f - s - m)))
)

constraints = [
    f - s >= m + eps,
    s >= 0,
    f >= 0,
    cp.max(f) <= 20
]

# precedence constraints
for (i, j) in P:
    constraints += [s[j] >= f[i]]

prob = cp.Problem(objective, constraints)
prob.solve()
```

```
Out[62]: np.float64(2.656825645230484)
```

```
In [65]: visualize_schedule(s.value, f.value)
```

