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Homework 1.1 Pooling problem

Recall the pooling problem from Exercise 1. In this homework, we consider another pooling problem instance presented in Figure 1 with 16 sources $s \in S$, 2 pools $p \in P$, and 2 target nodes $t \in T$. Let $N = S \cup P \cup T$ and let A denote the arcs between nodes in N .

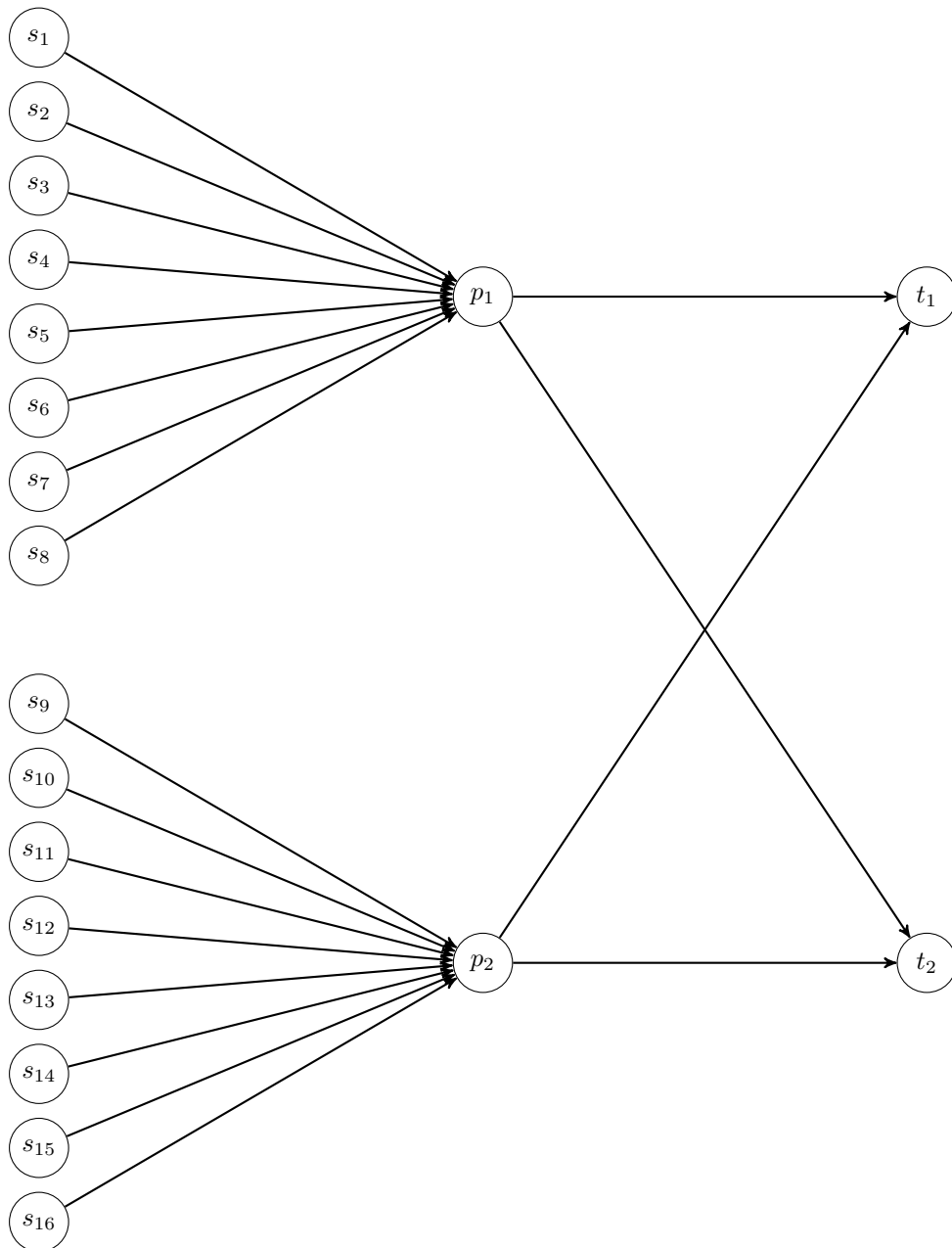


Figure 1: Pooling problem network.

The problem has 2 different properties q^1 and q^2 . Let $K = \{1, 2\}$ be the property index set. Each source node $s \in S$ has flow bounds and fixed property values. Each pool node $p \in P$ has flow bounds and unknown property values. Each target node has flow bounds and values bounds for both properties. The data for this problem is presented in Tables 1 and 2.

Table 1: Flow bounds and property values sources $s \in S$

$s \in S$	Flow bounds		Property values	
	\underline{b}_s	\bar{b}_s	\underline{q}_s^1	\bar{q}_s^2
s_1	1	3	5.84	43.7
s_2	0	3	5.40	36.8
s_3	0	3	0.24	12.8
s_4	0	3	2.01	15.4
s_5	1	3	5.85	47.3
s_6	0	3	5.38	39.2
s_7	0	3	0.26	13.1
s_8	0	3	2.04	15.9
s_9	1	3	0.64	39.9
s_{10}	0	3	0.57	38.2
s_{11}	0	3	0.02	13.5
s_{12}	0	3	0.14	16.3
s_{13}	1	3	0.93	38.1
s_{14}	0	3	0.85	34.1
s_{15}	0	3	0.03	13.2
s_{16}	0	3	0.26	15.5

Table 2: Flow bounds of pools $p \in P$ and flow and property bounds of targets $t \in T$

$p \in P$	Flow bounds		$t \in T$	Flow bounds		Property bounds			
	\underline{b}_p	\bar{b}_p		\underline{b}_t	\bar{b}_t	\underline{q}_t^1	\bar{q}_t^1	\underline{q}_t^2	\bar{q}_t^2
p_1	0	15	t_1	10	11	0	1.50	30.0	34.0
p_2	0	15	t_2	11	17	0	3.50	32.0	40.0

The source nodes have no costs in this example, and the costs c_t at target nodes $t \in T$ representing profit are nonlinear and based on the final values of property q^1 . The costs are defined as follows:

$$c_{t_1} = 100 \left(2 - \frac{q_{t_1}^1}{\bar{q}_{t_1}^1} \right) \quad \text{and} \quad c_{t_2} = 150 \left(2 - \frac{q_{t_2}^1}{\bar{q}_{t_2}^1} \right) \quad (1)$$

The problem can be formulated as follows.

$$\underset{x,q}{\text{maximize}} \sum_{t \in T} c_t \sum_{j \in N_t^-} x_{jt} \quad (2)$$

$$\text{subject to} \sum_{j \in N_s^+} x_{sj} \geq \underline{b}_s, \quad \forall s \in S \quad (3)$$

$$\sum_{j \in N_s^+} x_{sj} \leq \bar{b}_s, \quad \forall s \in S \quad (4)$$

$$\sum_{j \in N_p^-} x_{jp} = \sum_{j \in N_p^+} x_{pj}, \quad \forall p \in P \quad (5)$$

$$\sum_{j \in N_p^-} x_{jp} \leq \bar{b}_p, \quad \forall p \in P \quad (6)$$

$$\sum_{j \in N_t^-} x_{jt} \geq \bar{b}_t, \quad \forall t \in T \quad (7)$$

$$\sum_{j \in N_t^-} x_{jt} \leq \underline{b}_t, \quad \forall t \in T \quad (8)$$

$$\sum_{j \in N_p^-} q_j^k x_{jp} = q_p^k \sum_{j \in N_p^+} x_{pj}, \quad \forall p \in P, \forall k \in K \quad (9)$$

$$\sum_{j \in N_t^-} q_j^k x_{jt} = q_t^k \sum_{j \in N_t^-} x_{jt}, \quad \forall t \in T, \forall k \in K \quad (10)$$

$$q_t^k \geq \underline{q}_t^k, \quad \forall t \in T, \forall k \in K \quad (11)$$

$$q_t^k \leq \bar{q}_t^k, \quad \forall t \in T, \forall k \in K \quad (12)$$

$$q_s^k = \bar{q}_s^k, \quad \forall s \in S, \forall k \in K \quad (13)$$

$$q_p^k \geq 0, \quad \forall p \in P, \forall k \in K \quad (14)$$

$$x_{ij} \geq 0, \quad \forall (i,j) \in A \quad (15)$$

- Model and solve the problem (2) – (15) with Julia using JuMP using the data in Tables 1 and 2. Use the target costs defined in (1) in the objective. Compare with the pooling problem presented Exercise 1 if you get stuck.
- Try to find at least 2 different local optima by trying different values for the unknown quality variables q_p of the pool nodes $p \in P$.
- Remove all pool nodes $p \in P$ from the model and investigate how it affects the objective function values. Without pool nodes, the different crude oils sent from source nodes $s \in S$ are blended at the target nodes $t \in T$ only.
- Remember to return your Julia code file when submitting your solutions.**

Homework 1.2 Portfolio optimization

For this problem, use the data file [prices.csv](#) which contains daily prices of $N = \{1, \dots, n\}$ assets over a time period of $T = \{1, \dots, m\}$ days. Let $x_i \geq 0$ denote the (long) position of asset $i \in N$ in a portfolio throughout the time period. The positions $x = (x_1, \dots, x_n)$ in the portfolio are scaled to represent fractions of the total investment, that is,

$$\sum_{i \in N} x_i = 1$$

Let p_i^t denote the daily price of asset $i \in N$ for all $t \in T$, and let r_i^t be the relative daily return of asset $i \in N$ for all $t \in T \setminus \{m\}$. These are computed as

$$r_i^t = \frac{p_i^{t+1} - p_i^t}{p_i^t}, \quad \forall i \in N, \forall t \in T \setminus \{m\}$$

Let $\mu = (\mu_1, \dots, \mu_n)$ denote the expected relative returns of the assets N , and let $\Sigma \in \mathbb{R}^{n \times n}$ be the corresponding covariance matrix. Thus, the expected relative return and variance of the portfolio $x = (x_1, \dots, x_n)$ are $\mu^\top x$ and $x^\top \Sigma x$, respectively. You can compute the initial data using the following Julia code presented in Lecture 1:

```
# Read the daily prices data from a .csv file.
data = readcsv("prices.csv")

# First row has the names of the stocks
stocks = data[1, 1:end-2]

# Last two columns has data and US$ rate
prices = data[2:end, 1:end-2]

# Returns are calculated as (p(t+1) - p(t))/p(t)
returns = diff(prices) ./ prices[1:end-1, :]

# Number of days and stocks in data
T, n = size(returns)

# Calculates expected return and covariance
μ, Σ = mean(returns, 1), cov(returns)
```

Your task is to analyze the performance of the portfolio x by considering different trade-offs between the expected return $\mu^\top x$ and variance $x^\top \Sigma x$. The variance $x^\top \Sigma x$ measures exposition to risk, and ideally we would like to find a portfolio in which the expected return is large but the variance is small. These two conflicting objectives are combined in the following portfolio optimization problem proposed by Markowitz (1952):

$$\underset{x}{\text{maximize}} \quad \lambda(\mu^\top x) - (1 - \lambda)x^\top \Sigma x \quad (16)$$

$$\text{subject to} \quad \sum_{i \in N} x_i = 1 \quad (17)$$

$$x \geq 0 \quad (18)$$

The parameter λ with $0 \leq \lambda \leq 1$ scalarizes the two objectives into a single objective function (16). The value of λ depends on the investor's preference: to reduce risk of the investment, the investor would select a smaller value of λ . Conversely, a risk seeking investor with hopes of higher return would select a larger value of λ .

- Model and solve the problem (16) – (18) with Julia using Jump with different values of λ . Use the values $\lambda \in (0, 0.1, 0.2, \dots, 1)$. Discuss briefly how the return $\mu^\top x$ and risk $x^\top \Sigma x$ change with different values of λ .
- Plot the values of $\mu^\top x$ and $x^\top \Sigma x$ at the different points of λ from part (a). Plot both cases separately.
- Plot the values of $\mu^\top x$ at the corresponding points $x^\top \Sigma x$ using the results of part (a) obtained with different values of λ .
- Remember to return your Julia code file when submitting your solutions.**

References

Markowitz, H. (1952). Portfolio selection. *The journal of finance*, 7(1), 77–91.