

Chapter 5 *Topology design and analysis*

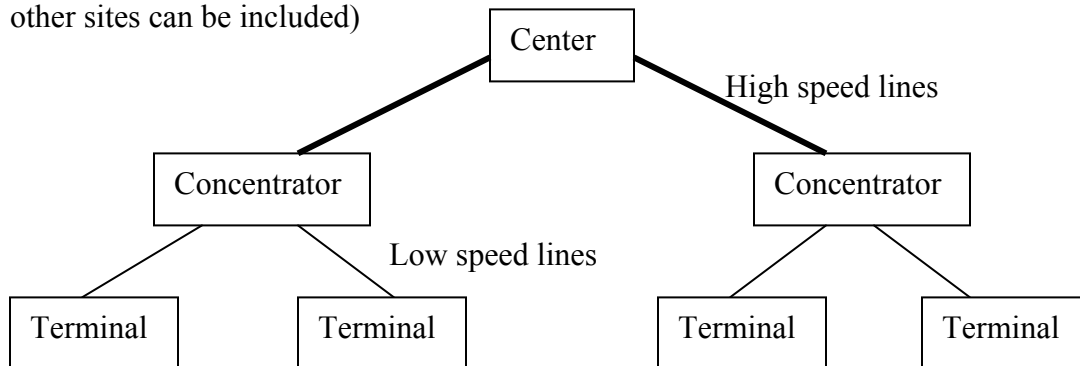
Topics covered:

Topology design. Network design algorithms. Terminal assignment. Concentrator location. Traffic flow analysis and performance evaluation. Network reliability. Network simulation.

5.1 Topology design

5.1.1 Centralized Network design

- **Centralized network:** is where all communication is to and from a single central site.
- The “central site” is capable of making routing decisions.
→ Tree topology provides only one path through the center (For reliability, lines between other sites can be included)



- Three different problems:
 - **Multipoint line topology:** selection of links connecting terminals to concentrators or directly to the center.
 - **Terminal assignment:** association of terminals with specific concentrators.
 - **Concentrator location:** deciding where to place concentrators, and whether or not to use them at all.

5.1.2 Finding Trees in Graphs

- Used to design and analyze networks.
- Connect a number of nodes to a central node:
 - **Node:** Hub, Switch, Router, etc.
 - **Central node:** backbone

- A tree is a graph with no loops, with only one path between any pair of nodes.
- Trees are minimal networks: provide connectivity without any unnecessary additional links:
 - Minimally reliable and robust
 - Networks are more highly connected (but design starts with a tree)

5.1.2.1 Tree Traversals

- Visit all nodes in a tree: edges are traversed twice.
- First, identify a node as the root
- Assume the tree is directed (outward from the root)
- Two algorithms:
 - BFS (Breadth First Search):
 - Nodes closest to root are visited first
 - Implemented using a queue (FIFO)
 - DFS (Depth First Search):
 - Visits an unvisited neighbor of the node just visited.
 - Implemented using a stack (LIFO)
- Both traversals (BFS and DFS) can be preorder traversals (i.e., visit nodes then successors) or post-order traversals (i.e., successors visited first).
- Traversal is generalized to undirected graphs by keeping track of which nodes were visited, and not visiting them again.
- In a BFS or DFS traversal, edges visited form a tree (if the graph is connected) or a forest (if the graph is not connected).

5.1.2.2 Minimum Spanning Trees (MSTs)

- Use DFS to find a spanning tree in a graph, if one exists
 - Arbitrary tree
- Useful to find the “best” tree
 - Minimum Spanning Tree (e.g., minimum total length. Where length is: distance, cost, function(delay), function(reliability), etc.)

- If the graph is not connected → minimum spanning forest
 - For n nodes, c components, and e edges, we have: $n = c + e$
 - For a tree, $c = 1$.
- DFS will not, in general, find the spanning tree with minimum total cost.

5.1.2.2.1 *The Greedy Algorithm*

- At each stage, select the shortest edge possible.
- May not find a feasible solution when one exists.
- Efficient and simple to implement → widely used.
- Basis of other more complex and effective algorithms.
- In the case of MST, the greedy algorithm guarantees both optimality and reasonable computational complexity.
 - Start with empty solution s
 - While elements exist
 - Find e , the best element not yet considered
 - If adding e to s is feasible, add it; if not, discard it.

5.1.2.2.2 *Kruskal's Algorithm*

- A greedy algorithm for finding MSTs.
- Sort the edges, shortest first and then include all edges which do not form cycles with the edges previously selected.
- n : number of nodes
- **Algorithm:**
 1. Sort all edges in ascending order (least cost first)
 2. Select among edges not yet selected, the one with the least cost.
 3. Add it if it does not create a cycle.
 4. If the number of edges selected $< n-1$, go to step (2), otherwise exit (tree completed)
- **Complexity:**

$O(m \log m)$, m = number of edges

5.1.2.2.3 Prim's Algorithm

- A greedy algorithm for finding MSTs.
- Advantageous if the network is dense.
- Well suited to parallel implementation.
- **Algorithm:**
 1. Start with one node (root node) in the tree
 2. Find node i , not in the tree, which is the nearest to the tree.
 3. Add node i to the tree and edge e connecting i to the tree.
- **Complexity:**

$$O(n^2)$$

5.1.2.2.4 Comparison of the Complexity of Kruskal's and Prim's Algorithms

- If the network is dense $\rightarrow m \sim O(n^2) \rightarrow$ Prim's algorithm is faster
- If the network is not dense $\rightarrow m \sim O(n) \rightarrow$ Kruskal' algorithm is faster

5.1.3 Constrained/Capacitated MST (CMST)

- The algorithms presented in the previous subsections are called “unconstrained MST algorithms”
 - No constraint on flow of information
 - No constraint on the number of ports at each node.
- For the unconstrained spanning tree problem, all these algorithms produce a minimum cost spanning tree.
- **CMST Problem:** Given a central node N_0 and a set of other nodes (N_1, N_2, \dots, N_n) , as et of weights (W_1, W_2, \dots, W_n) for each node, the capacity of a link, W_{\max} , and a cost matrix $C_{ij} = \text{Cost}(i,j)$, find a set of trees T_1, T_2, \dots, T_k such that each N_i belongs to exactly one T_j and each T_j contains N_0 .

- **Objective:** Find a tree of minimum cost and which satisfies a number of constraints such as:
 - Flow over a link
 - Number of ports

- **Example:**
 - Assume we are allowed to use one type of links only that can accommodate a maximum of 5 units of flow per unit time.
 - Assume that the flow generated from each node to the central node (N_1) is as follows: $f_1=0$, $f_2=2$, $f_3=3$, $f_4=2$, $f_5=1$ (in units/time_unit).

- Effect of constraint violation:
 - As a result, a queue will build up since node 3 can service only 5 units/time_unit. If node 3 does not have a large queue to accommodate all coming units, some units will be lost. So, these units are retransmitted, which may cause the network to collapse.

- The CSMT problem is NP-hard (i.e., cannot be solved in polynomial time)
 - Resort to heuristics (approximate algorithms)

- These heuristics will attempt to find a good feasible solution, not necessarily the best, that:
 - Minimizes the cost
 - Satisfies all the constraints

- Well-known heuristics:
 - Kruskal
 - Prim
 - Esau-Williams

5.1.3.1 Kruskal's Algorithm for CMST

Algorithm:

1. Sort all edges in ascending order, $e \leftarrow 0$.
2. Select edge with minimum cost (from edges not yet selected)
3. If it satisfies constraints (i.e., no cycles and no violation of flows on links)
 - o Then: add it to the tree, $e \leftarrow e + 1$
 - o Else: go to step (2)
4. If ($e = n - 1$) then exit, else go to step (2)

Example:

Given a network with five nodes, labelled **1** to **5**, and characterized by the following cost matrix:

	1	2	3	4	5
1	-	3	3	5	10
2	3	-	6	4	8
3	3	6	-	3	5
4	5	4	3	-	7
5	10	8	5	7	-

Node 1 is the central backbone node.

$$f_{\max}=5, f_1=0, f_2=2, f_3=3, f_4=2, f_5=1.$$

5.1.3.2 Prim's Algorithm for CMST

Algorithm:

1. Start with one node (root node) in the tree.
2. Find node i , not in the tree, which is the nearest to the tree
3. Add node i to the tree and edge e connecting i to the tree if it satisfies constraints (i.e., no violation of flows on links)

Example:

5.1.3.3 Esau-Williams Algorithm for CMST

Node 1 is the central node.

t_{ij} : is the tradeoff of connecting i to j or i directly to the root.

- If ($t_{ij} < 0$) → better to connect i to j
- If ($t_{ij} \geq 0$) → better to connect i directly to the root

Algorithm:

1. Compute $t_{ij} = c_{ij} - c_{i1}$ for all $i, j \neq 1$.
2. Select the link (m,n) such that: $t_{mn} = \min(t_{ij})$
3. If $t_{mn} < 0$, then go to step (4)
Else (i.e., $t_{mn} \geq 0$ for all m,n), connect to node 1 all nodes not connected yet, and **exit**.
4. Verify constraints (e.g., does not exceed the maximum weight)
 - If satisfied go to step (5)
 - Else: $t_{mn} = \infty$ and $t_{nm} = \infty$, go to step (2)
5. Add link (m, n) , remove link $(m, 1)$ and update t_{ij} to indicate that m is now connected to n .
 - $t_{mn} = \infty$ and $t_{nm} = \infty$
 - if $t_{mj} \neq \infty$, $t_{mj} = c_{mj} - \min(c_{ki})$ [$k \in C_m$, where C_i = component containing node i]
6. Go to step (2)

Example:

Given a network with five nodes, labelled 1 to 5, and characterized by the following cost matrix:

	1	2	3	4	5
1	-	3	3	5	10
2	3	-	6	4	8
3	3	6	-	3	5
4	5	4	3	-	7
5	10	8	5	7	-

$W_{\max}=5, W_1=0, W_2=2, W_3=3, W_4=2, W_5=1$.

5.1.4 Terminal Assignment

5.1.4.1 Problem Statement

- **Terminal Assignment:** Association of terminals with specific concentrators.

Given:

T terminals (stations) $i = 1, 2, \dots, T$

C Concentrators (hubs/switches) $j = 1, 2, \dots, C$

C_{ij} : cost of connecting terminal i to concentrator j

W_j : capacity of concentrator j

Assume that terminal i requires W_i units of a concentrator capacity.

Assume that the cost of all concentrators is the same.

- $x_{ij} = 1$; if terminal i is assigned to concentrator j .

- $x_{ij} = 0$; otherwise.

Objective:

5.1.4.2 Augmenting path algorithm

Based on the following observations:

1. Ideally, every terminal is assigned to the nearest concentrator.
2. Terminals on concentrators that are full are moved only to make room for another terminal that would cause a higher overall cost if assigned to another concentrator.
3. An optimal partial solution with $k+1$ terminals can be found by finding the least expensive way of adding the $(k+1)^{\text{th}}$ terminal to the k terminal solution.

Assignment problem:

Given a cost matrix:

- One column per concentrator
- One row per terminal

Assume that:

- Weight of each terminal is 1 (i.e., each terminal consumes exactly one unit of concentrator capacity)
- A concentrator has a capacity of W terminals (e.g., number of ports)

A feasible solution exists iff $T \leq W * C$

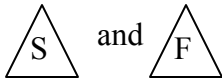
Algorithm:

1. Initially, try to associate each terminal to its nearest concentrator
2. If successful in assigning all terminals without violating capacity constraints, then stop (i.e., an optimal solution is found)
3. Else,
 - **Repeat**
 - i. Build a compressed auxiliary graph
 - ii. Find an optimal augmentation
 - **Until** all terminals are assigned

Building a compressed auxiliary graph:

U: set of unassociated terminals

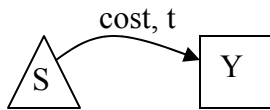
T(Y): set of terminals associated with Y



are the start and finish of all augmenting paths

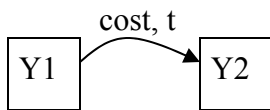


represents a fully loaded concentrator



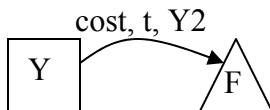
Assign t to a fully loaded concentrator Y. ($t \in U$)

$\text{cost} = c(tY) = \min c(xY)$ for $x \in U$



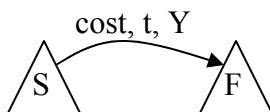
Move t from a fully loaded concentrator Y1 to another fully loaded concentrator Y2. ($t \in T(Y1)$)

$\text{cost} = c(tY2) - c(tY1) = \min (c(xY2) - c(xY1))$ for $x \in t(Y1)$



Move t from Y to a concentrator Y2 with spare capacity. ($t \in T(Y)$)

$\text{cost} = c(tY2) - c(tY) = \min (c(xY2) - c(xY))$ for $x \in t(Y)$



Assign t to a concentrator Y with spare capacity. ($t \in U$)

$\text{cost} = c(tY) = \min c(xY)$ for $x \in U$

Example:

5.2 Traffic Flow Analysis and Performance Evaluation

5.2.1 Traffic Flow Analysis Objective

- Estimate:
 - Delay
 - Utilization of resources (links)
- Traffic flow across a network depends on:
 - Topology
 - Routing
 - Traffic workload (from all traffic sources)
- Desirable topology and routing are associated with:
 - Low delays
 - Reasonable link utilization (no bottlenecks)
- Assumptions:
 - Topology is fixed and stable
 - Links and routers are 100% reliable
 - Processing time at the routers is negligible
 - Capacity of all links is given $C = [C_i]$ (in bps [bits per second])
 - Traffic workload is given $\Gamma = [\gamma_{jk}]$ (in pps [packets per second])
 - Routing is given $R = [r_{jk}]$
 - Average packet size is $1/\mu$ bits.

5.2.2 Queuing Analysis

Projections of performance are made on the basis of either:

- The existing load information, or
- The estimated load for the new environment.

Approaches that could be used:

- Do an after-the-fact analysis based on actual values
- Make a simple projection from existing to expected environment
- Develop an analytic model based on queuing theory
- Program and run a simulation tool

5.2.2.1 Queuing Models

- The notation **X/Y/N** is used for queuing models.
 - X = distribution of the interarrival times
 - Y = distribution of service times
 - N = number of servers
- The most common distributions are:
 - G = general independent arrivals or service times
 - M = negative exponential distribution
 - D = deterministic arrivals or fixed length service

➤ Example: **M/M/1**

➤ **Single-server queues**

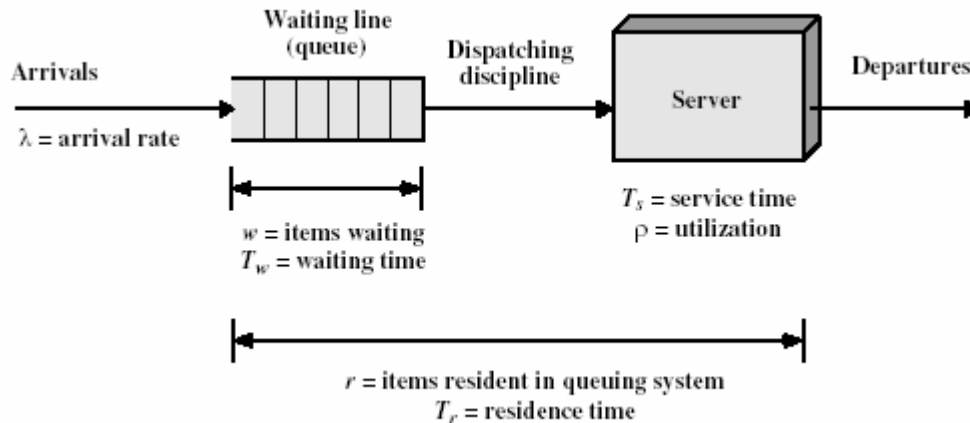


Figure 5.1: Queuing System Structure and Parameters for Single-Server Queue
(Taken from "Queuing Analysis" by William Stallings)

Queue parameters:

λ = arrival rate; mean number of arrivals per second

T_s = mean service time for each arrival; amount of time being served, not counting time waiting in the queue

ρ = utilization; fraction of time facility (server or servers) is busy

r = mean number of items in system, waiting and being served (residence time)

T_r = mean time an item spends in system (residence time)

w = mean number of items waiting to be served

T_w = mean waiting time (including items that have to wait and items with waiting time = 0)

Basic Queuing relationship:

- $\rho = \lambda * T_s$
- $r = w + \rho$
- $\lambda_{\max} = 1/T_s$
- $r = \lambda * T_r$ (**Little's formula**)
- $w = \lambda * T_w$ (Little's formula)
- $T_r = T_w + T_s$
- $r = \rho / (1 - \rho)$

➤ **Multiserver queue**

N = number of servers

ρ = utilization of each server

$N\rho$ = utilization of all servers ($= \lambda * T_s$)

5.2.2.2 M/M/1 Queues – Application to Networks

- Each link is seen as a service station servicing packets.

λ_i = arrival rate (in pps); mean number of packets that arrive to link i in one second.

μC_i = average service rate (in pps); mean number of packets that will get out of the link i in one second. ($= 1/T_s$)

- Utilization of link i is:

$$\rho_i =$$

- Stability condition of a network is:

- The external workload offered to the network is:

$$\gamma =$$

Where:

- γ = total workload in packets per second
- γ_{jk} = workload between source j and destination k
- N = total number of sources and destinations

- The internal workload on link i is:

$$\lambda_i =$$

Where:

- γ_{jk} = workload between source j and destination k
- Π_{jk} = path followed by packets to go from source j and destination k

- The total internal workload is:

$$\lambda =$$

Where:

- λ = total load on all of the links in the network
- λ_i = load on link i
- L = total number of links

- The average length for all paths is given by:

$$\mathbf{E[\text{number of links in a path}] = \lambda/\gamma}$$

- The average number of items waiting and being served for link i is:

$$\mathbf{r_i =}$$

- The number of packets waiting and being served in the network can be expressed as:

$$\gamma * T =$$

Where:

T = average delay experienced by a packet through the network.

$$T =$$

- T_{ri} is the residence time at each queue. If we assume that each queue can be treated as an independent M/M/1 model (Jackson's Theorem), then:

$$T_{ri} =$$

Where: T_{si} is the service time for link i

$$T_{si} =$$

Where:

- C_i = data rate on the link (in bps)
- $M = 1/\mu$ = average packet length in bits

Example:

5.3 Network Reliability

5.3.1 Introduction

- A network model is a set of facilities. A facility could be a device or a link.
- A network must contain some slack to allow it to function even if some of its facilities have failed.
- Any network facility is either:
 - Working (**p**)
 - Failing ($q = 1-p$)
- **MTBF**: Mean Time Between Failures (**f**).
- **MTTR**: Mean Time To Repair (**r**)

- For any facility i , we'll know from measurements of f_i and r_i :

$$P_i = \text{Prob} [\text{facility } i \text{ is working}] =$$

Therefore:

- We assume that all facilities are independent:

$$P(\mathbf{ij}) = \text{Prob}[\text{facility } i \text{ and facility } j \text{ are working}] =$$

$$P(\mathbf{i|j}) = \text{Prob}[\text{facility } i \text{ or facility } j \text{ is working}] =$$

- Simplest measure of network reliability:

$$P_c(\mathbf{G}) = \text{Prob}[\text{Network is connected}]$$

Where: **c** stands for the connectivity of the network, and
G stands for the graph representing the network

$P_c(G) = \text{Prob}[\text{All nodes are working and there is a spanning tree of working links}]$

$P_c(G) =$

- Since enumerating all trees in G requires an exponential amount of effort, $P_c(G)$ is very difficult (if not impossible) to compute.

→ We seek simpler measures of network reliability.

5.3.2 Reliability of Tree Networks

- A typical enterprise/campus network includes trees:

- Given a tree T :

**$P_c(T) = \text{Prob}[\text{A tree network, } T, \text{ being connected}]$
 $= \text{Prob}[\text{All components (nodes and links) are working}]$**

$P_c(T) =$

- $P_c(T)$ can also be computed recursively:

$P_c(T) =$

Where: **$T-i$** is the tree T without node i , and
 j is the link between node i and the rest of the tree

- Given a particular tree with root r:

$$P_c(\mathbf{i}) = \text{Prob}[\text{node } i \text{ can communicate with root } r]$$

$$P_c(\mathbf{i}) =$$

Where: \mathbf{j} is the link between nodes i and k , and
 \mathbf{k} is the predecessor of node i

$$P_c(\mathbf{r}) =$$

- The expected number of nodes communicating with the root r is:

$$E(\mathbf{r}) =$$

- This expression can be computed efficiently for any node as follows:

$E(\mathbf{i})$ = the expected number of nodes communicating with the node i

$$E(\mathbf{i}) =$$

- If node i is a leaf, then:

$$E(\mathbf{i}) =$$

Example:

- The expected number of node pairs communicating through the root r is:

$$EPR(r) =$$

Example:

5.4 References

1. “Telecommunications Network Design Algorithms” by Aaron Kershenbaum, 1993
2. “Queuing Analysis” by William Stalling, 2000