

# Routing Algorithms

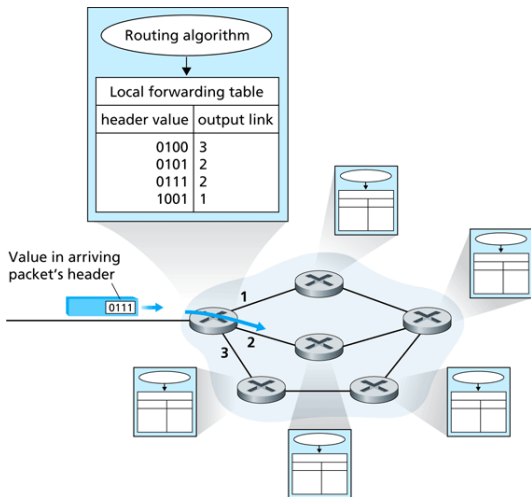
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# Routing

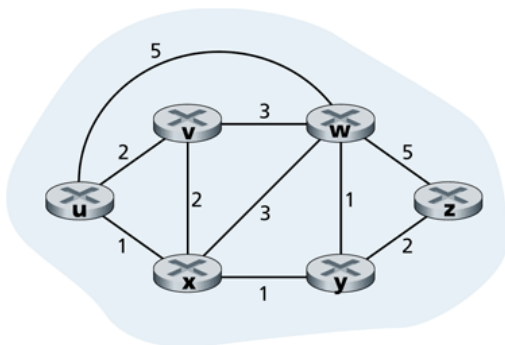
- **How does the Internet determine which path to use from the source to the destination?**
- Challenges
  - need to handle hundreds of thousands of routes
  - need to handle changes anywhere in the network
  - stability is crucial

# Packet Forwarding versus Routing



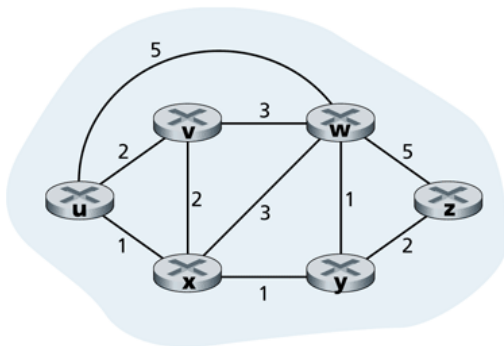
# Graph Abstraction

- graph  $G = (V, E)$
- nodes/vertices  $V = u, v, w, x, y, z$
- links/edges  $E =$   
 $(u, v), (u, x), (v, x), (v, w), (x, w), (x, y), (w, y), (w, z), (y, z)$
- useful for virtual networks as well, e.g. to model a peer-to-peer network of TCP connections



# Edge Weights or Costs

- edges can have weights/costs
  - $c(w, z) = 5$
  - costs may be all 1
- cost of a path is the sum of the cost of all edges in the path
- routing algorithm finds the least-cost path**



# Types of Routing Algorithms

- link-state algorithm
  - routers advertise their links to every other router
  - eventually, all routers know complete topology and link costs
- distance-vector algorithm
  - routers start knowing path to immediate neighbors
  - routers advertise all known destinations and path lengths to neighbors
  - eventually, all routers know which neighbor has shortest path to all destinations

# Link-State Routing

# Link-State Basics

- can use Dijkstra's algorithm to compute least cost paths from one source to all other nodes
- notation
  - $c(x, y)$ : link cost from node  $x$  to  $y$ ,  $\infty$  if no link between  $x$  and  $y$
  - $D(v)$ : current cost of path from source to destination  $v$
  - $F$ : set of nodes whose least cost path definitively known



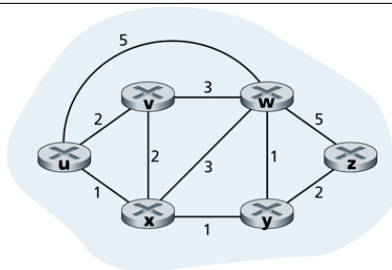
# Dijkstra's Algorithm

```
F = [u]
for v in N:
    if v adjacent to u:
         $D(v) = c(u, v)$ 
    else:
         $D(v) = \text{infinity}$ 
while N != F:
    find w not in F such that  $D(w)$  is a minimum
    F.append(w)
    update  $D(v)$  for all v adjacent to w and not in F
         $D(v) = \min(D(v), D(w) + c(w, v))$ 
```

# Example: Routing Table for U

- $h(v)$ : next hop on shortest path toward  $v$

Step	F	$D(v), h(v)$	$D(w), h(w)$	$D(x), h(x)$	$D(y), h(y)$	$D(z), h(z)$
0	u	2, u	5, u	1, u	$\infty$	$\infty$
1	ux	2, u	4, x		2, x	$\infty$
2	uxy	2, u	3, y		4, y	
3	uxyv		3, y		4, y	
4	uxyvw				4, y	
5	uxyvwz					



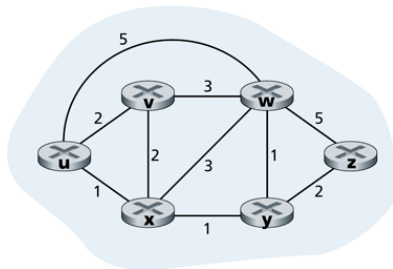
# Algorithm Complexity

- each iteration needs to check all nodes  $w$  not in  $F$
- $N(N + 1)/2$  comparisons
- $O(N^2)$
- more efficient implementations possible:  $O(N \log N)$

# Distance-Vector Routing

# Bellman-Ford

- distributed algorithm
  - $d_x(y)$ : cost of least-cost path from  $x$  to  $y$
  - $d_x(y) = \min(c(x, v) + d_v(y))$
  - min is taken over all neighbors of  $x$
- example
  - $d_v(z) = 5, d_x(z) = 3, d_w(z) = 3$
  - $d_u(z) = \min(c(u, v) + d_v(z), c(u, x) + d_x(z), c(u, w) + d_w(z))$
  - $d_u(z) = \min(2 + 5, 1 + 3, 5 + 3) = 4$
- node with minimum cost to destination is chosen as next hop



# Bellman-Ford

- start with known costs o neighbors
- calculate best estimate of  $d_x(y)$
- distance vector  $D_x = \{D_x(y) : y \in N\}$
- send distance vector to neighbors whenever it changes
- update  $D_x$  using Bellman-Ford rule whenever local link cost change or neighbor's vector results in a new minimum
- **distributed, asynchronous algorithm**

Diagram illustrating the construction of a Huffman tree using the greedy algorithm. The diagram shows three stages of the process, each with a cost matrix and a tree structure.

**Stage 1:** The cost matrix is:

	x	y	z
x	0	2	7
y	$\infty$	$\infty$	$\infty$
z	$\infty$	$\infty$	$\infty$

The tree structure shows node x (cost 0) and node y (cost 2) being merged into a new node (cost 2). Node z (cost 7) remains separate.

**Stage 2:** The cost matrix is:

	x	y	z
x	0	2	3
y	2	0	1
z	7	1	0

The tree structure shows node x (cost 0) and node y (cost 2) being merged into a new node (cost 2). Node z (cost 7) remains separate.

**Stage 3:** The cost matrix is:

	x	y	z
x	0	2	3
y	2	0	1
z	3	1	0

The tree structure shows node x (cost 0) and node y (cost 2) being merged into a new node (cost 2). Node z (cost 7) remains separate.

	cost to				cost to				cost to		
	x	y	z		x	y	z		x	y	z
from x	∞	∞	∞	from x	0	2	7	from x	0	2	3
y	2	0	1	y	2	0	1	y	2	0	1
z	∞	∞	∞	z	7	1	0	z	3	1	0

from

	cost to		
	x	y	z
x	∞	∞	∞
y	∞	∞	∞
z	7	1	0

from

	cost to		
	x	y	z
x	0	2	7
y	2	0	1
z	3	1	0

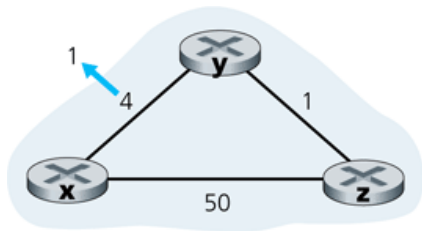
from

	cost to		
	x	y	z
x	0	2	3
y	2	0	1
z	3	1	0

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# Reduction in Link Cost

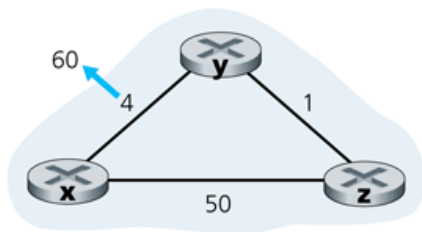
- good news travels fast
  - at  $t_1$ , y detects cost change, updates DV, informs neighbors
  - at  $t_2$ , z receives update from y, updates its DV, informs neighbors
  - at  $t_3$ , y receives z's update, DV does not change, process finishes





# Increase in Link Cost

- bad news travels slow
  - at  $t_0$ ,  
 $d_y(x) = 4$ ,  $d_z(x) = 5$
  - at  $t_1$ , y detects link change, sets  $d_y(x) = c(y, z) + d_z(x) = 6$ , sends DV to z
  - at  $t_2$ , z gets DV, updates  $d_z(x) = c(z, y) + d_y(x) = 7$
  - loop continues until  
 $d_y(x) = 60$
- **count to infinity problem**



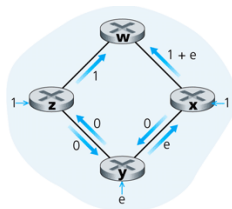
# Solutions To Count To Infinity

- poisoned reverse
  - if  $z$ 's best path is through  $y$ ,  $z$  tells  $y$  that  $d_z(x)$  is infinite
  - prevents  $y$  from routing to  $x$  via  $z$
  - doesn't solve all looping problems
- path vector routing
  - include a path with each route, not just the next hop
  - eliminates all count-to-infinity problems
  - optimization: include only the next-to-last hop

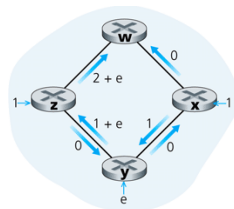
# Load-Based Routing

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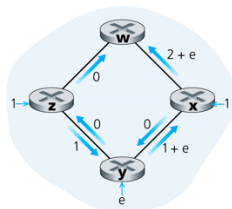
- Difficult to have stable routes when link costs reflect congestion
- e.g. cost = number of traffic flows



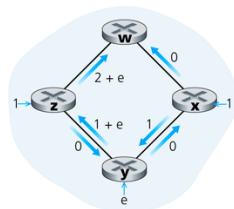
a. Initial routing



b. x, y detect better path to w, clockwise



c. x, y, z detect better path to w, counterclockwise



d. x, y, z detect better path to w, clockwise