



INDIAN INSTITUTE OF SCIENCE

# **Water Resources Systems:** **Modeling Techniques and Analysis**

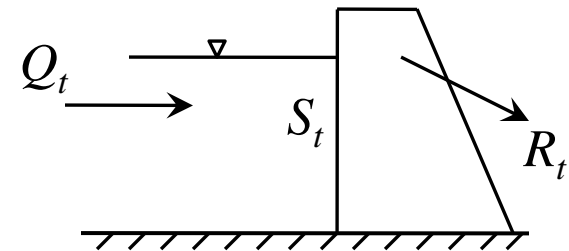
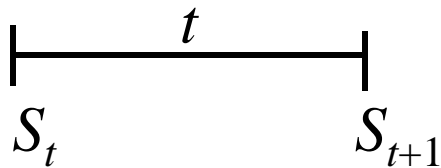
Lecture - 17

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# Summary of the previous lecture

- Reservoir operation problem



Storage continuity (Mass balance)

$$S_{t+1} = S_t + Q_t - R_t \quad \text{..... Neglecting losses}$$

where  $S_t$  : Storage at the beginning of period  $t$

$Q_t$  : Inflow during period  $t$

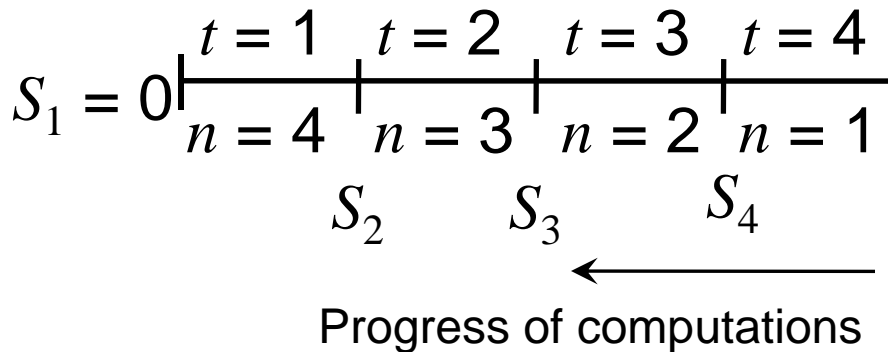
$R_t$  : Release during period  $t$

$$f_t^n (S_t) = \underset{\substack{0 \leq R_t \leq (S_t + Q_t) \\ S_t + Q_t - R_t \leq S_{\max}}}{\text{Max}} \left[ B_t (R_t) + f_{t+1}^{n-1} (S_t + Q_t - R_t) \right]$$

# Dynamic Programming

Example:

- Inflows during four seasons to a reservoir with storage capacity of 4 units are 2, 1, 3 and 2 units respectively.



Release	Benefits
0	-100
1	250
2	320
3	480
4	520
5	520
6	410
7	120

# Dynamic Programming

Stage 1:

$$Q_4 = 2 \quad t = 4 \quad \text{and} \quad n = 1$$

$$f_4^1(S_4) = \text{Max} [B_4(R_4)]$$

$$0 \leq R_4 \leq (S_4 + Q_4)$$

$$S_4 + Q_4 - R_4 \leq 4$$

# Dynamic Programming

$$Q_4 = 2$$

$S_4$	$R_4$	$B_4(R_4)$	$f_4^1(S_4) = \text{Max}[B_4(R_4)]$	$R_4^*$
0	0	-100	320	2
	1	250		
	2	320		
1	0	-100	480	3
	1	250		
	2	320		
	3	480		
2	0	-100	520	4
	1	250		
	2	320		
	3	480		
	4	520		

Contd.

# Dynamic Programming

$Q_4 = 2$   
Contd.

$S_4$	$R_4$	$B_4(R_4)$	$f_4^1(S_4) = \text{Max}[B_4(R_4)]$	$R_4^*$
3	1	250	520	4,5
	2	320		
	3	480		
	4	520		
	5	520		
4	2	320	520	4,5
	3	480		
	4	520		
	5	520		
	6	410		

# Dynamic Programming

Stage 2:

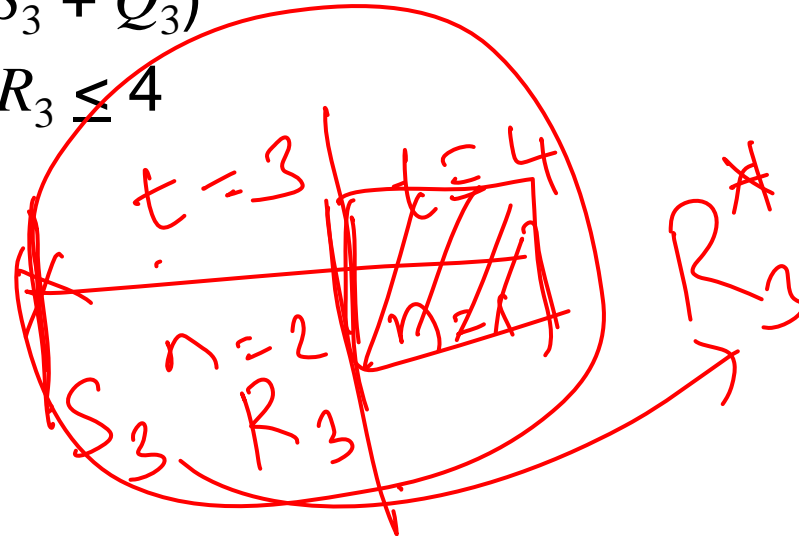
$$Q_3 = 3$$

$$t = 3 \text{ and } n = 2$$

$$f_3^2(S_3) = \text{Max} [B_3(R_3) + f_4^1(S_3 + Q_3 - R_3)]$$

$$0 \leq R_3 \leq (S_3 + Q_3)$$

$$S_3 + Q_3 - R_3 \leq 4$$



# Dynamic Programming

$$Q_3 = 3$$

$$f_3^2(S_3) = \text{Max} [B_3(R_3) + f_4^1(S_3 + Q_3 - R_3)]$$

$$0 \leq R_3 \leq (S_3 + Q_3) ; S_3 + Q_3 - R_3 \leq 4$$

$S_3$	$R_3$	$B_3(R_3)$	$S_3 + Q_3 - R_3$	$f_4^1(S_3 + Q_3 - R_3)$	$B_3(R_3) + f_4^1(S_3 + Q_3 - R_3)$	$f_3^2(S_3)$	$R_3^*$
0	0	-100	3	520	420	800	2, 3
	1	250	2	520	770		
	2	320	1	480	800		
	3	480	0	320	800		
1	0	-100	4	520	420	960	3
	1	250	3	520	770		
	2	320	2	520	840		
	3	480	1	480	960		
	4	520	0	320	840		

Contd.



Contd.  $Q_3 = 3$

$S_3$	$R_3$	$B_3(R_3)$	$S_3 + Q_3 - R_3$	$f_4^1(S_3 + Q_3 - R_3)$	$B_3(R_3) + f_4^1(S_3 + Q_3 - R_3)$	$f_3^2(S_3)$	$R_3^*$
2	1	250	4	520	770	1000	3, 4
	2	320	3	520	840		
	3	480	2	520	1000		
	4	520	1	480	1000		
	5	520	0	320	840		
3	2	320	4	520	840	1040	4
	3	480	3	520	1000		
	4	520	2	520	1040		
	5	520	1	480	1000		
	6	410	0	320	730		
4	3	480	4	520	1000	1040	4, 5
	4	520	3	520	1040		
	5	520	2	520	1040		
	6	410	1	480	890		
	7	120	0	320	440		

# Dynamic Programming

Stage 3:

$$Q_2 = 1$$

$$t = 2 \text{ and } n = 3$$

$$f_2^3(S_2) = \text{Max} \left[ B_2(R_2) + f_3^2(S_2 + Q_2 - R_2) \right]$$

$$0 \leq R_2 \leq (S_2 + Q_2)$$

$$S_2 + Q_2 - R_2 \leq 4$$

*from previous stage table*

# Dynamic Programming

$$f_2^3(S_2) = \text{Max} [B_2(R_2) + f_3^2(S_2 + Q_2 - R_2)]$$

$$0 \leq R_2 \leq (S_2 + Q_2)$$

$$S_2 + Q_2 - R_2 \leq 4$$

$$Q_2 = 1$$

$S_2$	$R_2$	$B_2(R_2)$	$S_2 + Q_2 - R_2$	$f_3^2(S_2 + Q_2 - R_2)$	$B_2(R_2) + f_3^2(S_2 + Q_2 - R_2)$	$f_2^3(S_2)$	$R_2^*$
0	0	-100	1	960	860	1050	1
	1	250	0	800	1050		
1	0	-100	2	1000	900	1210	1
	1	250	1	960	1210		
	2	320	0	800	1120		
2	0	-100	3	1040	940	1280	2, 3
	1	250	2	1000	1250		
	2	320	1	960	1280		
	3	480	0	800	1280		

Contd. 11

$Q_2 = 1$

# Dynamic Programming

Contd.

$S_2$	$R_2$	$B_2(R_2)$	$S_2 + Q_2 - R_2$	$f_3^2(S_2 + Q_2 - R_2)$	$B_2(R_2) + f_3^2(S_2 + Q_2 - R_2)$	$f_2^3(S_2)$	$R_2^*$
3	0	-100	4	1040	940	1440	3
	1	250	3	1040	1290		
	2	320	2	1000	1320		
	3	480	1	960	1440		
	4	520	0	800	1320		
4	1	250	4	1040	1290	1480	3, 4
	2	320	3	1040	1360		
	3	480	2	1000	1480		
	4	520	1	960	1480		
	5	520	0	800	1320		

# Dynamic Programming

Stage 4:

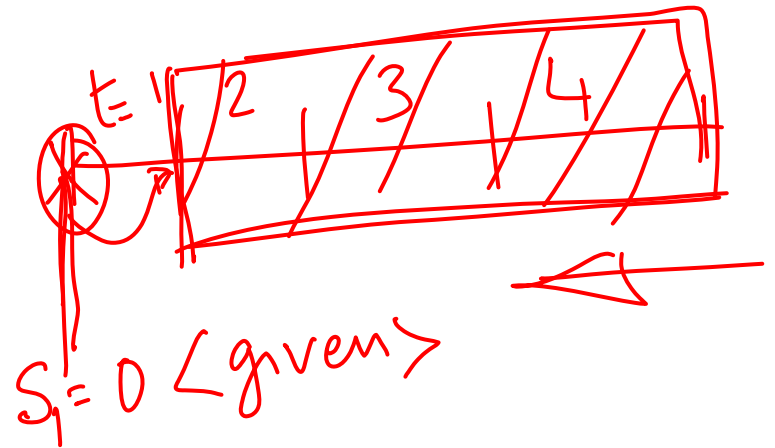
$$Q_1 = 2$$

$$t = 1 \quad \text{and} \quad n = 4$$

$$f_1^4(S_1) = \text{Max} \left[ B_1(R_1) + f_2^3(S_1 + Q_1 - R_1) \right]$$

$$0 \leq R_1 \leq (S_1 + Q_1)$$

$$S_1 + Q_1 - R_1 \leq 4$$



# Dynamic Programming

$$f_1^4(S_1) = \text{Max} [B_1(R_1) + f_2^3(S_1 + Q_1 - R_1)]$$

$$0 \leq R_1 \leq (S_1 + Q_1)$$

$$S_1 + Q_1 - R_1 \leq 4$$

$$Q_1 = 2$$

$S_1$	$R_1$	$B_1(R_1)$	$S_1 + Q_1 - R_1$	$f_2^3(S_1 + Q_1 - R_1)$	$B_1(R_1) + f_2^3(S_1 + Q_1 - R_1)$	$f_1^4(S_1)$	$R_1^*$
0	0	-100	2	1280	1180	1460	1
	1	250	1	1210	1460		
	2	320	0	1050	1370		

Handwritten red annotations:  $S_2$  with an arrow pointing to the  $f_2^3$  column, and a box containing  $t=1$  and numbers 2, 3, 4, 1.

# Dynamic Programming

Trace back:

$R_1^* = 1$  ..... From last table

$$\begin{aligned} S_2 &= S_1 + Q_1 - R_1^* \\ &= 0 + 2 - 1 = 1 \end{aligned}$$

$R_2^* = 1$  ..... From stage-3 table, corresponding to  $S_2 = 1$

$$\begin{aligned} S_3 &= S_2 + Q_2 - R_2^* \\ &= 1 + 1 - 1 = 1 \end{aligned}$$

$R_3^* = 3$  ..... From stage-2 table, corresponding to  $S_3 = 1$

# Dynamic Programming

$$\begin{aligned} S_4 &= S_3 + Q_3 - R_3^* \\ &= 1 + 3 - 3 = 1 \end{aligned}$$

$R_4^* = 3$  ..... From stage-1 table, corresponding to  $S_4 = 1$

- The optimal release sequence for the problem is  $\{1, 1, 3, 3\}$  during the four periods.
- The maximum net benefits that result from this release policy is 1460 units.



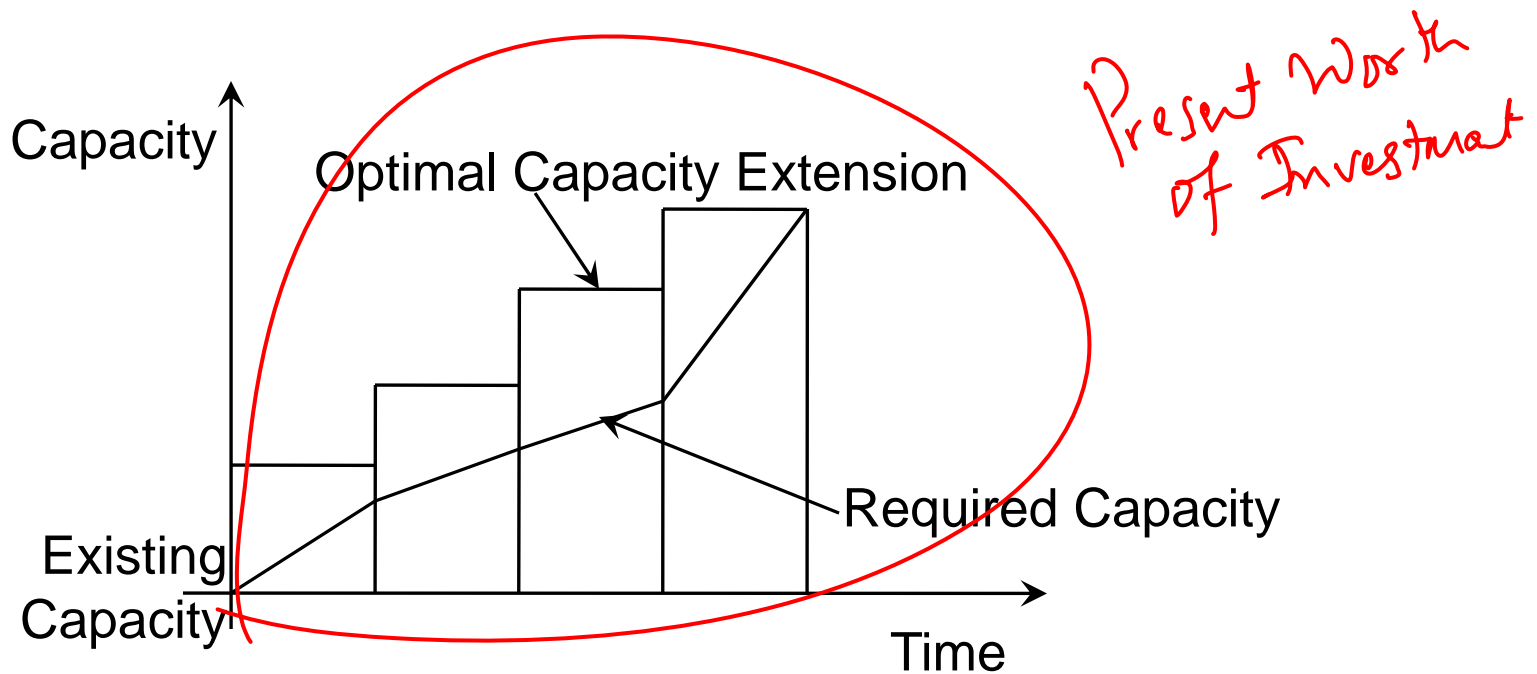
# Dynamic Programming

Capacity expansion problem:

- Expansion of existing capacities of an entire system or individual components of a system is often required.
- The decisions deal with investments at different times, starting with the present time, on capacity expansion needed over years in future.
- A typical problem in this class of problems is to decide in what steps the expansion over the next  $n$  years should be carried out so that the present worth of the investment is a minimum.

# Dynamic Programming

- For example we may need to decide on the investment on capacity expansion during next 5 years, 5 years after that and so on, if the planned capacity is to be achieved after, say, 25 years.



# Dynamic Programming

Example:

- A city water supply project envisages expansion of the storage system from the existing capacity of 100 units to 200 units in the next 20 years.
- The additional capacity required at the end of each of the 5 years is as follows

Time	Required additional capacity (units)
End of 5 <sup>th</sup> year	20
End of 10 <sup>th</sup> year	40
End of 15 <sup>th</sup> year	60
End of 20 <sup>th</sup> year	100

# Dynamic Programming

- The discounted present worth for additional capacities is as given below

t	Period (Years)	Additional capacity					
		0	20	40	60	80	100
		Discounted present worth of cost (units)					
1	1 – 5	0	120	150	200	250	280
2	6 – 10	0	80	110	130	150	
3	11 – 15	0	60	80	100		
4	15 – 20	0	40	50			

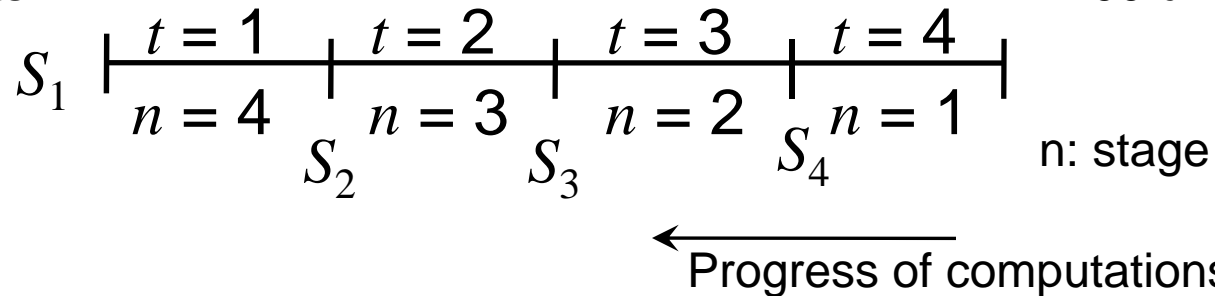
# Dynamic Programming

- Since the minimum additional capacity created at the end of the first 5 years is 20, only a maximum of 80 additional units need to be created during the remaining periods.
- The cost of additional capacity is therefore shown only for a maximum of 80 units for years 6 – 10.
- Similarly, for the other years, costs are shown only up to the maximum capacity that still needs to be created till the end of the planning horizon.

# Dynamic Programming

Existing capacity  
= 100 units

Desired capacity  
= 200 units



- Stage : period during which a decision on capacity expansion is required.
- State variable : capacity at the beginning of a period.
- Decision variable : capacity expansion during a period.

# Dynamic Programming

Stage 1:  $t = 4$

$$f_4(S_4) = \text{Min} [C_4(x_4)]$$

$$160 \leq S_4 \leq 200$$

$$S_4 + x_4 = 200$$

$S_4$	$x_4$	$C_4(x_4)$	$f_4(S_4) = \text{Min} [C_4(x_4)]$	$x_4^*$
160	40	50	50	40
180	20	40	40	20
200	0	0	0	0

# Dynamic Programming

Stage 2:  $t = 3$

$$f_3 (S_3) = \text{Min} [C_3 (x_3) + f_4 (S_3 + x_3)]$$

$$140 \leq S_3 \leq 200$$

$$160 \leq S_3 + x_3 \leq 200$$



# Dynamic Programming

$$f_3(S_3) = \text{Min} [C_3(x_3) + f_4(S_3 + x_3)]$$

$S_3$	$x_3$	$C_3(S_3)$	$S_3+x_3$	$f_4(C_3+S_3)$	$C_3(S_3)+f_4(C_3+S_3)$	$f_3(S_3)$	$x_3^*$
140	20	60	160	50	110	100	60
	40	80	180	40	120		
	60	100	200	0	100		
160	0	0	160	50	50	50	0
	20	60	180	40	100		
	40	80	200	0	80		
180	0	0	180	40	40	40	0
	20	60	200	0	60		
200	0	0	200	0	0	0	0

# Dynamic Programming

Stage 3:  $t = 2$

$$f_2(S_2) = \text{Min} [C_2(x_2) + f_3(S_2 + x_2)]$$

$$120 \leq S_2 \leq 200$$

$$140 \leq S_2 + x_2 \leq 200$$

$S_2$	$x_2$	$C_2(S_2)$	$S_2+x_2$	$f_3(C_2+S_2)$	$C_2(S_2)+$ $f_3(C_2+S_2)$	$f_2(S_2)$	$x_2^*$
120	20	80	140	100	180	150	80
	40	110	160	50	160		
	60	130	180	40	170		
	80	150	200	0	150		

# Dynamic Programming

$$f_2(S_2) = \text{Min} [C_2(x_2) + f_3(S_2 + x_2)]$$

$S_2$	$x_2$	$C_2(S_2)$	$S_2+x_2$	$f_3(C_2+S_2)$	$C_2(S_2)+f_3(C_2+S_2)$	$f_2(S_2)$	$x_2^*$
140	0	0	140	100	100	100	0
	20	80	160	50	130		
	40	110	180	40	150		
	60	130	200	0	130		
160	0	0	160	50	50	50	0
	20	80	180	40	120		
	40	110	200	0	110		
180	0	0	180	40	40	40	0
	20	80	200	0	80		
200	0	0	200	0	0	0	0

# Dynamic Programming

Stage 4:  $t = 1$

$$f_1(S_1) = \text{Min} [C_1(x_1) + f_2(S_1 + x_1)]$$

$$S_1 = 100$$

$$120 \leq S_1 + x_1 \leq 200$$

$S_1$	$x_1$	$C_1(S_1)$	$S_1+x_1$	$f_2(C_1+S_1)$	$C_1(S_1)+$ $f_2(C_1+S_1)$	$f_1(S_1)$	$x_1^*$
100	20	120	120	150	270	250	40, 60
	40	150	140	100	250		
	60	200	160	50	250		
	80	250	180	40	290		
	100	280	200	0	280		

# Dynamic Programming

Trace back of the solution:

	$x_1^*$	$x_2^*$	$x_3^*$	$x_4^*$	
	40	0	60	0	
Capacity:	100	140	140	200	200

Alternate solution is:

	$x_1^*$	$x_2^*$	$x_3^*$	$x_4^*$	
	60	0	0	40	
Capacity:	100	160	160	160	200

Optimum present worth of the investment = 250 units.


# Dynamic Programming

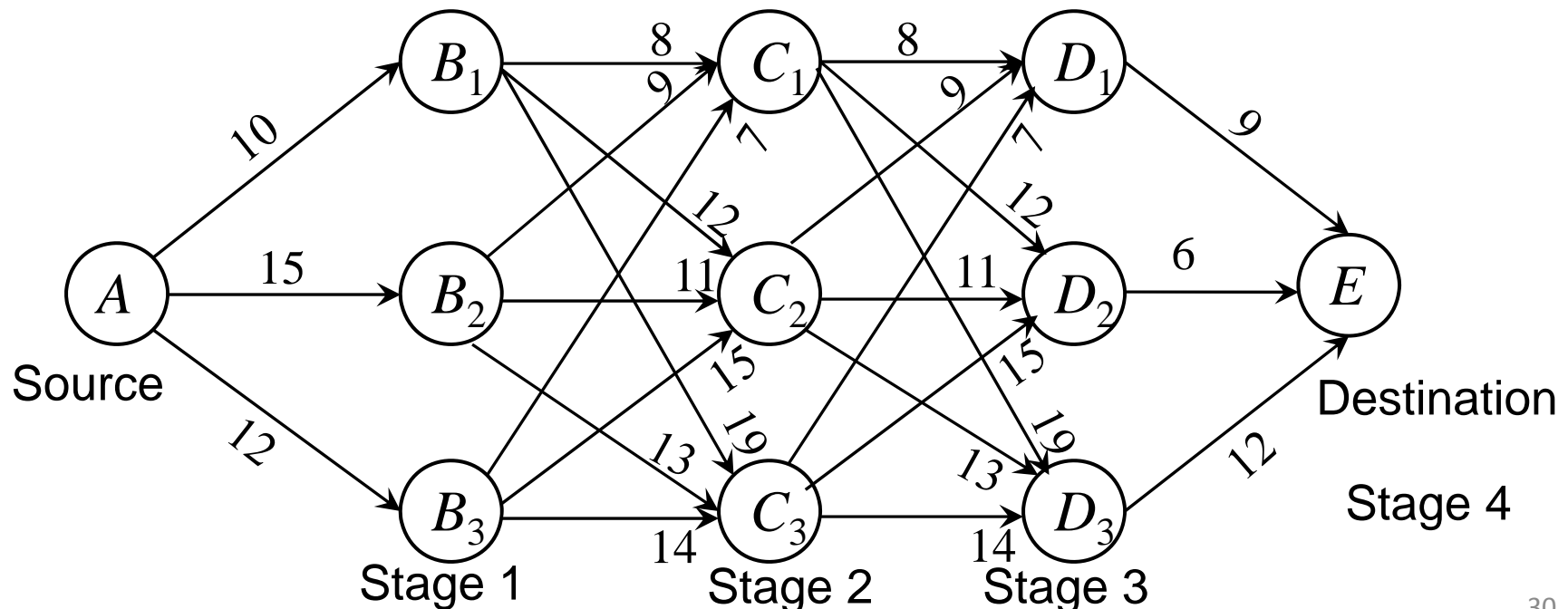
Shortest route problem:

- Consider the problem of determining the shortest route for a pipeline from among various possible routes available from destination to source.

State  $S_n$ : node in stage  $n$

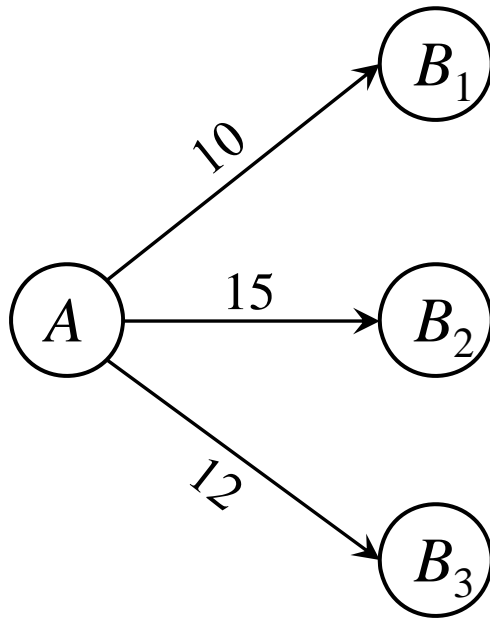
Decision  $x_n$ : node in the previous stage  $n-1$  from which the node  $S_n$  is reached.

Progress of computations  




# Dynamic Programming

Stage 1:



$$f_1^*(S_1) = \text{Min} [d(x_1, S_1)]$$

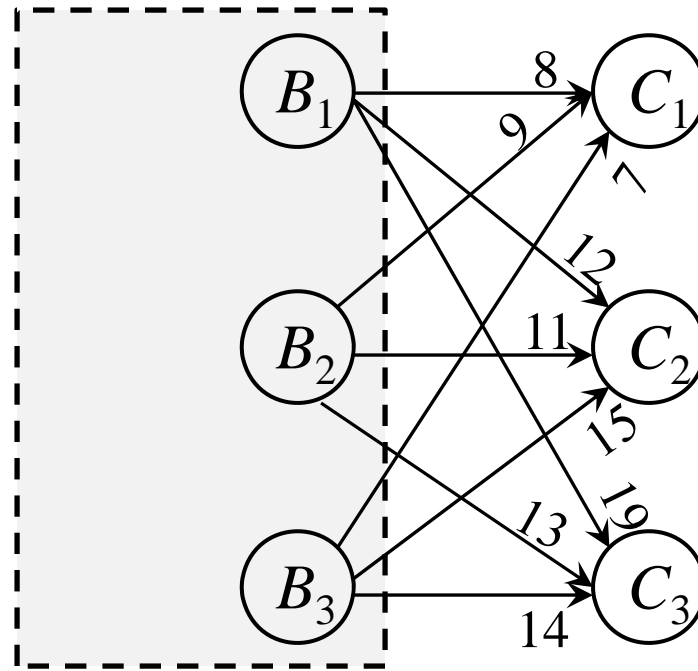
# Dynamic Programming

$S_1$	$x_1$	$f_1^*(S_1) = \text{Min} [d(x_1, S_1)]$	$x_1^*$
$B_1$	10	10	A
$B_2$	15	15	A
$B_3$	12	12	A



# Dynamic Programming

Stage 2:



$$f_2^*(S_2) = \text{Min} \left[ d(x_2, S_2) + f_1^*(x_2) \right]$$

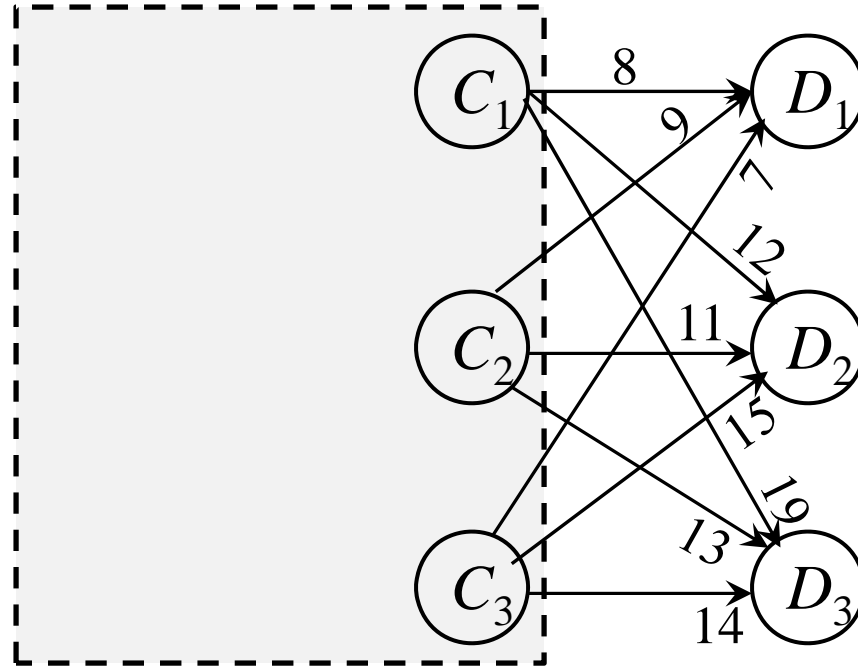
# Dynamic Programming

$$f_2^*(S_2) = \text{Min} [d(x_2, S_2) + f_1^*(x_2)]$$

$S_2$	$x_2$	$d(x_2, S_2)$	$f_1^*(x_2)$	$d(x_2, S_2) + f_1^*(x_2)$	$f_2^*(S_2)$	$x_2^*$
$C_1$	$B_1$	8	10	18	18	$B_1$
	$B_2$	9	15	24		
	$B_3$	7	12	19		
$C_2$	$B_1$	12	10	22	22	$B_1$
	$B_2$	11	15	26		
	$B_3$	15	12	27		
$C_3$	$B_1$	19	10	29	26	$B_3$
	$B_2$	13	15	28		
	$B_3$	14	12	26		

# Dynamic Programming

Stage 3:



$$f_3^*(S_3) = \text{Min} \left[ d(x_3, S_3) + f_2^*(x_3) \right]$$

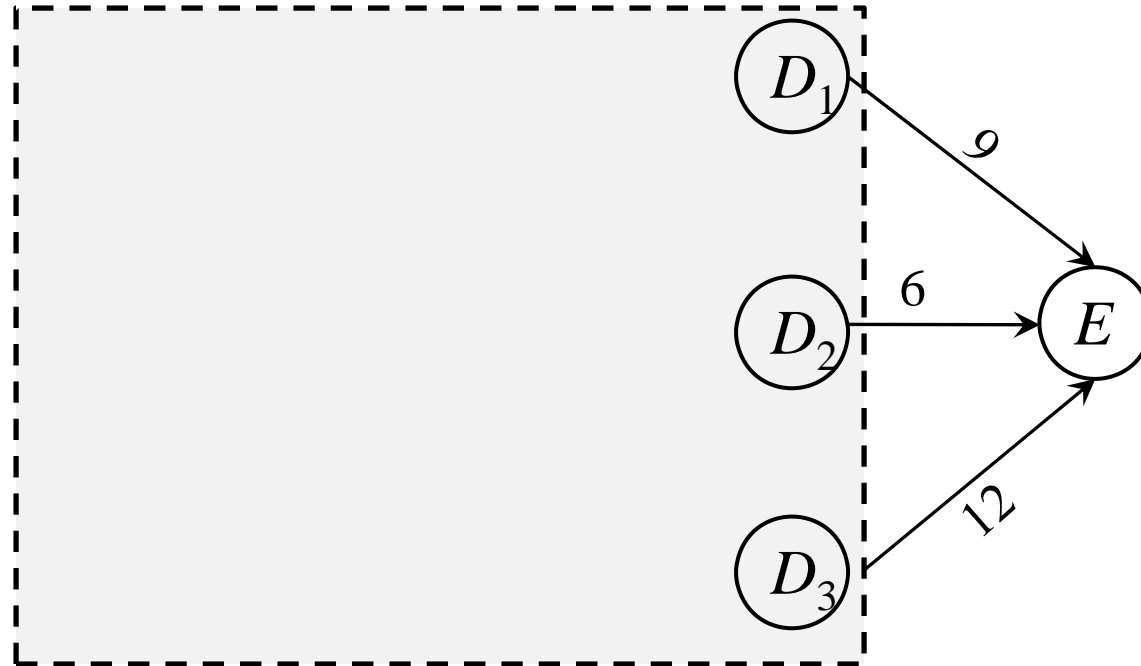
# Dynamic Programming

$$f_3^*(S_3) = \text{Min} [d(x_3, S_3) + f_2^*(x_3)]$$

$S_3$	$x_3$	$d(x_3, S_3)$	$f_2^*(x_3)$	$d(x_3, S_3) + f_2^*(x_3)$	$f_3^*(S_3)$	$x_3^*$
$D_1$	$C_1$	8	18	26	26	$C_1$
	$C_2$	9	22	29		
	$C_3$	7	26	33		
$D_2$	$C_1$	12	18	30	30	$C_1$
	$C_2$	11	22	33		
	$C_3$	15	26	41		
$D_3$	$C_1$	19	18	37	35	$C_2$
	$C_2$	13	22	35		
	$C_3$	14	26	40		

# Dynamic Programming

Stage 4:



$$f_4^*(S_4) = \text{Min} \left[ d(x_4, S_4) + f_3^*(x_4) \right]$$

# Dynamic Programming

$$f_4^*(S_4) = \text{Min} [d(x_4, S_4) + f_3^*(x_4)]$$

$S_4$	$x_4$	$d(x_4, S_4)$	$f_3^*(x_4)$	$d(x_4, S_4) + f_3^*(x_4)$	$f_4^*(S_4)$	$x_4^*$
$E$	$D_1$	9	26	35	35	$D_1$
	$D_2$	6	30	36		
	$D_3$	12	35	47		

# Dynamic Programming

Trace back:

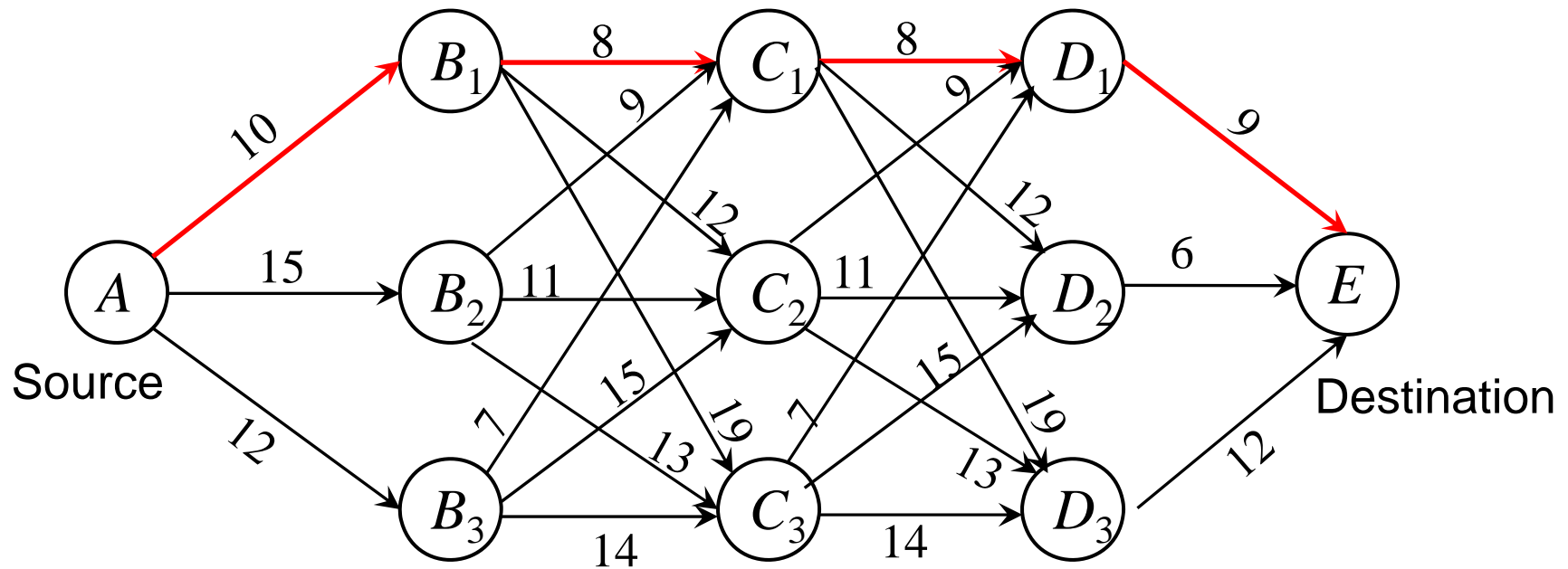
$x_4^* = D_1$  ..... From last table

$x_3^* = C_1$  ..... From stage-3 table, corresponding to  
 $S_3 = D_1$

$x_2^* = B_1$  ..... From stage-2 table, corresponding to  
 $S_2 = C_1$

$x_1^* = A$  ..... From stage-1 table, corresponding to  
 $S_1 = B_1$

# Dynamic Programming



Shortest distance = 35 units