

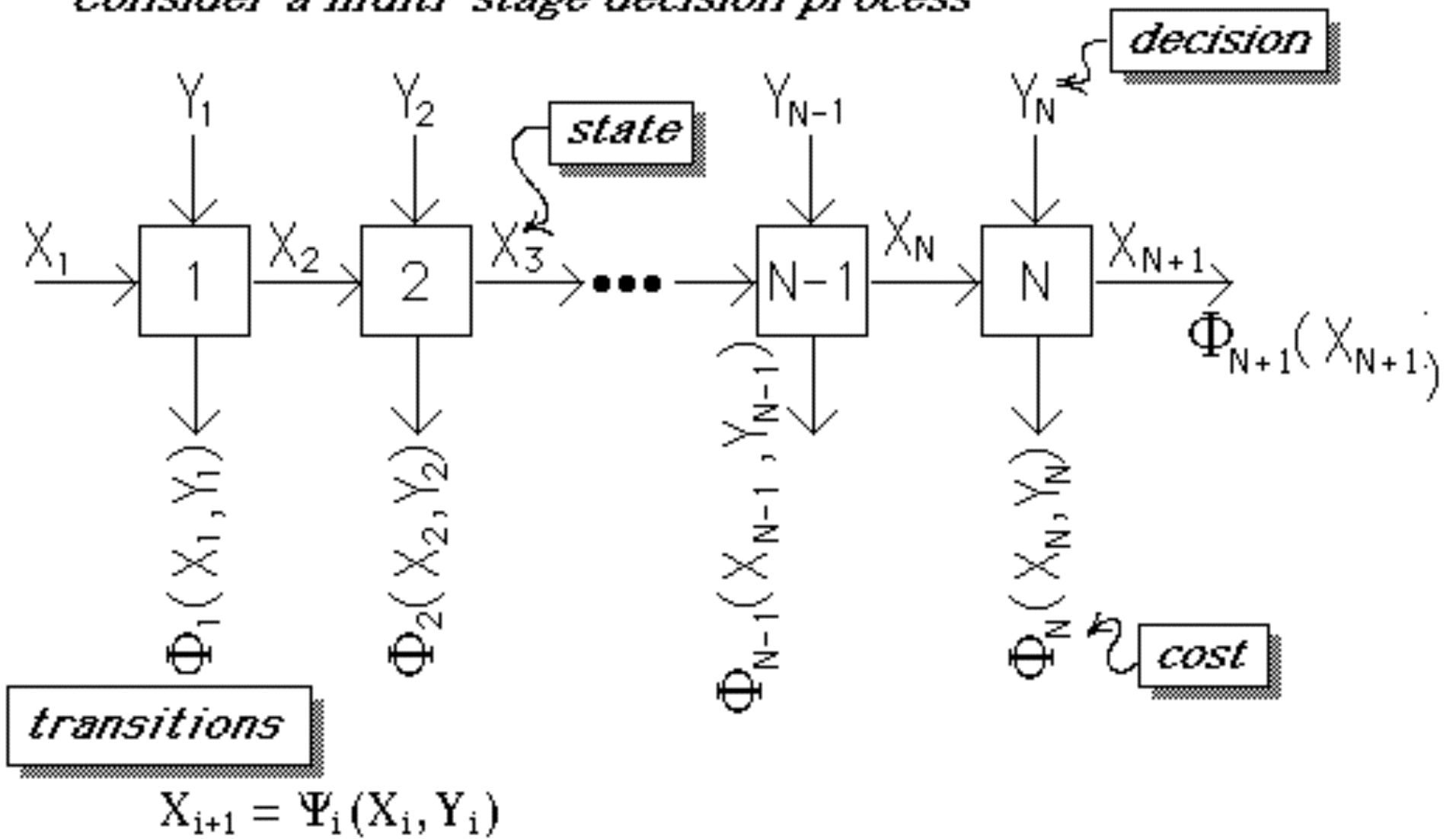
# **Successive Approximation Method**

## **Quadratic Criterion & Linear Dynamics**

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*Consider a multi-stage decision process*



$$\text{Minimize } \sum_{i=1}^N \Phi_i(X_i, Y_i)$$

subject to  $X_{i+1} = \Psi_i(X_i, Y_i), \quad i=1, 2, \dots, N$

If the objective function  $\Phi_i$  is **quadratic** in  $X$  and  $Y$ , and the transition function  $\Psi_i$  is **linear** in  $X$  and  $Y$  (the "QC/LD" case), we have a closed-form solution for the problem.

Otherwise, we can try **successively approximating** the problem by a QC/LD problem.

*(Successive approximation of a convex constrained programming problem by a linearly-constrained quadratic programming problem has been a very successful approach in nonlinear programming.)*

## Successive Approximation Algorithm (SAM)

Step 0: "Guess" at a decision sequence  $\bar{Y}_1, \bar{Y}_2, \dots, \bar{Y}_N$

Step 1: Use the transition functions together with the initial state  $X_1$  to compute a trajectory  $\bar{X}_1, \bar{X}_2, \bar{X}_3, \dots, \bar{X}_N$

Step 2: Compute  $\Phi_i(\bar{X}_i, \bar{Y}_i)$ , its gradient  $\nabla \Phi_i(\bar{X}_i, \bar{Y}_i)$

$$\text{i.e., } \frac{\partial}{\partial X_i} \Phi_i(\bar{X}_i, \bar{Y}_i), \frac{\partial}{\partial Y_i} \Phi_i(\bar{X}_i, \bar{Y}_i)$$

and its Hessian matrix  $\nabla^2 \Phi_i(\bar{X}_i, \bar{Y}_i)$ , i.e.,

$$\frac{\partial^2}{\partial X_i^2} \Phi_i(\bar{X}_i, \bar{Y}_i), \frac{\partial^2}{\partial X_i \partial Y_i} \Phi_i(\bar{X}_i, \bar{Y}_i), \text{ \& } \frac{\partial^2}{\partial Y_i^2} \Phi_i(\bar{X}_i, \bar{Y}_i)$$

*(SAM, continued)*

Step 3: Approximate the cost at each stage by the *Taylor series* expansion about up to & including the quadratic terms:

$$\begin{aligned} \Phi_i (X_i, Y_i) \approx & \Phi_i (\bar{X}_i, \bar{Y}_i) + \frac{\partial \Phi}{\partial X_i} (\bar{X}_i, \bar{Y}_i) (X_i - \bar{X}_i) + \frac{\partial \Phi}{\partial Y_i} (\bar{X}_i, \bar{Y}_i) (Y_i - \bar{Y}_i) \\ & + \frac{1}{2} \left[ \frac{\partial^2 \Phi}{\partial X_i^2} (\bar{X}_i, \bar{Y}_i) (X_i - \bar{X}_i)^2 + \frac{\partial^2}{\partial X_i \partial Y_i} \Phi_i (\bar{X}_i, \bar{Y}_i) (X_i - \bar{X}_i) (Y_i - \bar{Y}_i) \right. \\ & \left. + \frac{\partial^2 \Phi}{\partial Y_i^2} (\bar{X}_i, \bar{Y}_i) (Y_i - \bar{Y}_i)^2 \right] \end{aligned}$$

*(SAM, continued)*

Step 4: Approximate the transition function by a linear function:

$$X_{i+1} = \Psi_i (X_i, Y_i) \approx \Psi_i (\bar{X}_i, \bar{Y}_i) + \frac{\partial \Psi}{\partial X_i}(\bar{X}_i, \bar{Y}_i) (X_i - \bar{X}_i) + \frac{\partial \Psi}{\partial Y_i}(\bar{X}_i, \bar{Y}_i) (Y_i - \bar{Y}_i)$$

*(SAM, continued)*

Step 5: Use the closed-form solution to the QC/LD problem to compute the optimal decisions

$$\hat{Y}_1, \hat{Y}_2, \hat{Y}_3, \dots, \hat{Y}_N$$

Step 6: Use the transition functions, together with initial state  $X_1$  and decisions  $\hat{Y}_1, \hat{Y}_2, \hat{Y}_3, \dots, \hat{Y}_N$  to compute the new trajectory

$$\hat{X}_{i+1} = \Psi_i(\hat{X}_i, \hat{Y}_i) \quad , i=1,2,3,\dots,N$$

Step 7: If the termination criterion is not satisfied, let  $\bar{X} = \hat{X}$  and  $\bar{Y} = \hat{Y}$ , and return to step 2.

## EXAMPLE

$$\text{Minimize } Y_1^4 + X_2^4 + Y_2^4 + X_3^4$$

subject to

$$\begin{cases} X_{i+1} = X_i^2 + 4Y_i, & i=1,2 \\ X_1 = 2 \end{cases}$$

$$\bar{Y}_1 = -1, \quad \bar{Y}_2 = -0.1$$



## *APL Code for the Criterion Function*

Number of stages: N = 2

Objective Function

```
Z←T FN XY
```

```
⌘
```

```
⌘
```

Objective Function for SAMDP

```
⌘
```

```
Z←(0 1 1)[T]×XY[1]★4
```

```
→0 IF N<T
```

```
Z←Z+(1 1 0)[T]×XY[2]★4
```

## *APL Code for Gradient of Criterion Function*

```
G←T GRADF XY
R
R      Gradient of objective function for example problem
R
G←(0 1 1)[T]×4×XY[1]★3
→Last IF N<T    ◇ G←G,0    ◇ →0
Last:G←G,(1 1 0)[T]×4×XY[2]★3
```

## *APL Code for Hessian Matrix of Criterion Function*

```
H←T HESSIAN XY
R
R      Hessian matrix for objective function
R      of example problem
R
H←2 2ρ0
H[1;1]←(0 1 1)[T]×12×XY[1]★2
→0 IF N<T
H[2;2]←(1 1 0)[T]×12×XY[2]★2
```

## *APL Code for Transition Function*

```
Z←T PHI XY
```

```
R
```

```
R      Transition function for example problem
```

```
R
```

```
Z←(XY[1]*2) + 4*XY[2]
```

## *APL Code for Gradient of Transition Function*

```
G←T GRADPHI XY
```

```
R
```

```
R      Gradient of transition function for example problem
```

```
R
```

```
G←(2*XY[1]),4
```

## ITERATION #1

QC/LD Approximation at:

i	1	2	3
X[i]	2	0	-0.4
Y[i]	-1	-0.1	0

Objective function value is 1.0257

T	A	B	C	D	E	F	G	H	K
0	0	0	6	0	12	7	4	4	-4
1	0	0	0.06	0	0.012	0.0007	0	4	0
2	0.96	0	0	0.512	0	0.0768	0	0	0

Solution of QC/LD Approximation is:

i	1	2	3
X[i]	2	0	-0.267185
Y[i]	-1	-0.0667964	0

Sum of  $|Y - Y'| = 0.0332036$

## Iteration 2

QC/LD Approximation at:

i	1	2	3
X[i]	2	0	-0.267185
Y[i]	-1	-0.0667964	0

Objective function value is 1.00512

	A	B	C	D	E	F	G	H	K
0		0 6		0	12	7		4 4	-4
0		0 0.02677		0	0.003576	0.0001393		0 4	0
0.428328	0 0			0.152591	0	0.0152888		0 0	0

Solution of QC/LD Approximation is:

i	1	2	3
X[i]	2	0	-0.17847
Y[i]	-1	-0.0446175	0

Sum of |Y-Y'| = 0.0221788

QC/LD Approximation at:

Iteration 3

X(i) 2 0 -0.17847

Y(i)-1 -0.0446175 0

Objective function value is 1.00102

T	A	B	C	D	E
0	0E0	0	6E0	0E0	1.2E1
1	0E0	0	1.19444E-2	0E0	1.06586E-3
2	1.91111E-1	0	0E0	4.54765E-2	0E0

F	G	H	K
7E0	4	4	-4
2.77409E-5	0	4	0
3.04358E-3	0	0	0

Solution of QC/LD Approximation is:

i 1 2 3  
X(i) 2 0 -0.119212  
Y(i)-1 -0.0298029 0

Sum of |Y-Y'| = 0.0148146



QC/LD Approximation at:

Iteration 4

X[i] 2 0 -0.119212  
Y[i] 1 -0.0298029 0

Objective function value is 1.0002

T	A	B	C	D	E
0	0E0	0	6E0	0E0	1.2E1
1	0E0	0	5.32928E-3	0E0	3.17656E-4
2	8.52684E-2	0	0E0	1.35533E-2	0E0

	F	G	H	K
	7E0	4	4	-4
	5.52246E-6	0	4	0
	6.05892E-4	0	0	0

Solution of QC/LD Approximation is:

i 1 2 3  
X[i] 2 0 -0.079629  
Y[i] 1 -0.0199073 0

Sum of |Y-Y'| = 0.00989565

Since  $\sum_i |Y - Y'|$  is less than the tolerance (0.01), the algorithm terminates.