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**INFLUENCE DIAGRAM MODELS  
WITH CONTINUOUS VARIABLES**

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# INTRODUCTION



## OUTLINE OF PRESENTATION

- THEORY
  - INFLUENCE DIAGRAMS
  - NORMAL INFLUENCE DIAGRAM
- APPLICATIONS
  - DISCRETE - TIME FILTERING
  - $\Delta V$  - ACCURACY ANALYSIS

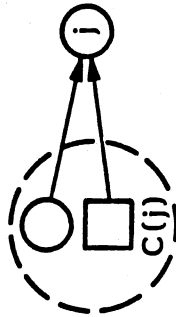
# INFLUENCE DIAGRAMS



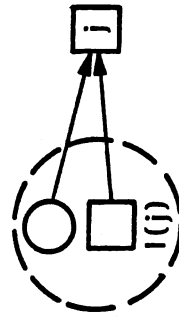
## DEFINITIONS

- ACYCLIC DIRECTED GRAPH WITH NODES  $N = \{1, \dots, n\}$
- VECTOR OF VARIABLES  $X = (X_1, \dots, X_n)^T = X_N$
- NODES
  - = CHANCE NODE IN  $C \subseteq N$  (RANDOM VARIABLES)
  - = DECISION NODE IN  $D \subseteq N$  (DECISION VARIABLES)
  - ◇ = VALUE NODE (CONDITIONAL EXPECTATION OF VALUE FUNCTION)

- ARCS



PROBABILISTIC CONDITIONING OF  $X_j$  ON  $X_{C(j)}$

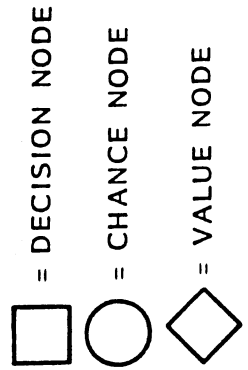
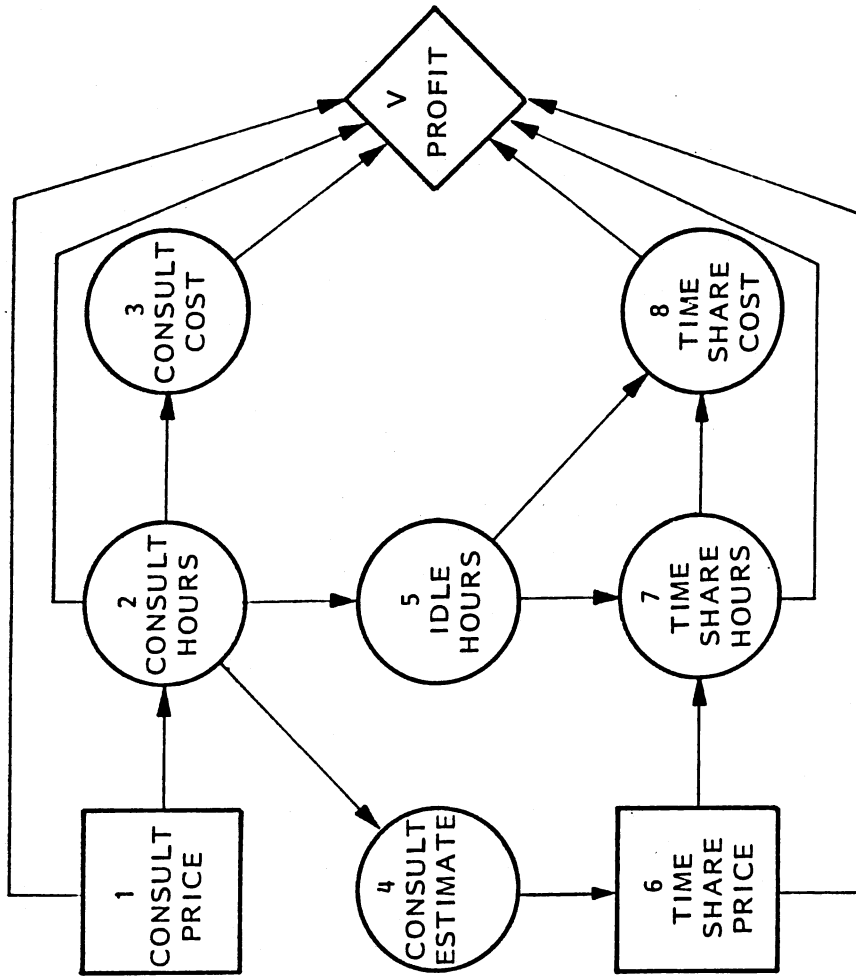


INFORMATION OBSERVABLE IS  $X_{I(j)}$  WHEN SELECTING DECISION  $X_j$



# INFLUENCE DIAGRAMS

## CONSULTANT'S PROBLEM

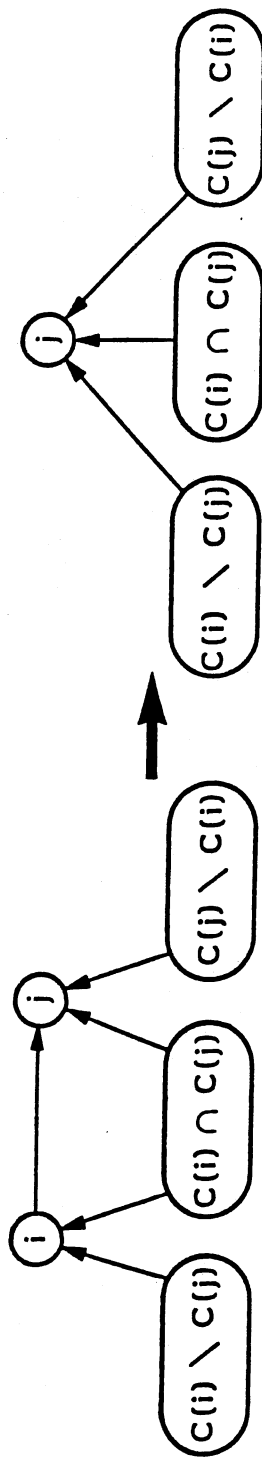




# INFLUENCE DIAGRAMS



## REMOVAL OF CHANGE NODE $i$ INTO CHANGE NODE $j$



BEFORE REMOVAL

AFTER REMOVAL

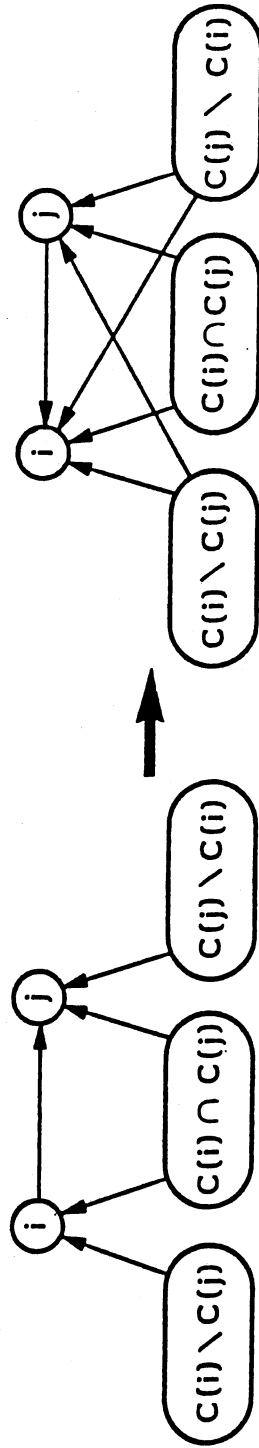
- " $\setminus$ " = SET SUBTRACTION
- INTEGRATION OF JOINT PROBABILITY WITH RESPECT TO R.V.  $X_i$
- NODE  $i$  MUST HAVE SINGLE SUCCESSOR NODE  $j$



# INFLUENCE DIAGRAMS



REVERSAL OF ARC FROM CHANCE NODE  $i$  TO CHANCE NODE  $j$



BEFORE REVERSAL

AFTER REVERSAL

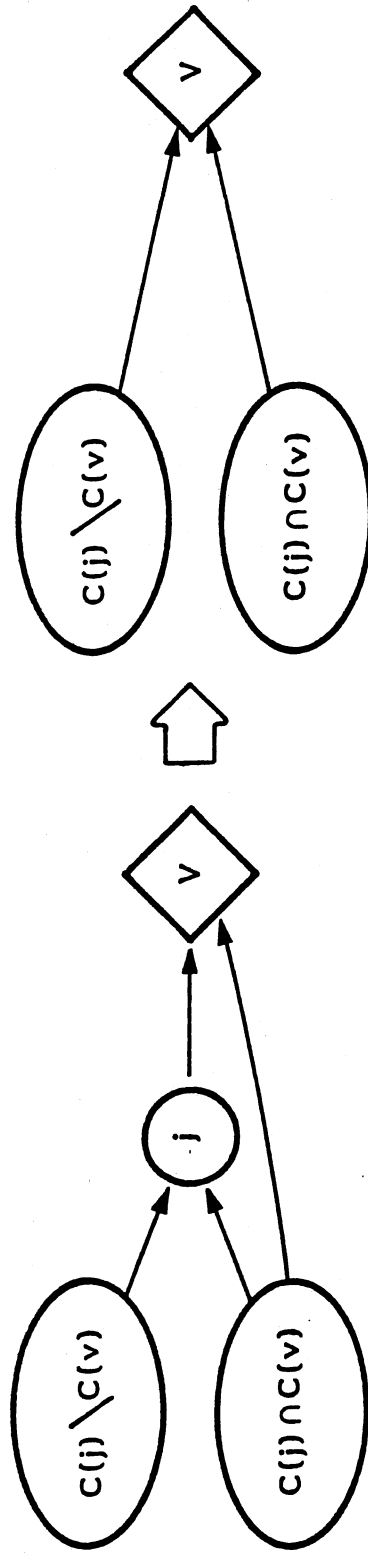
- INFLUENCE DIAGRAM VERSION OF BAYES' RULE
- REVERSAL CANNOT CREATE CYCLE (NO OTHER PATH FROM  $i$  TO  $j$ )
- NODES  $i$  AND  $j$  HAVE SAME CONDITIONING SET,  $C(i) \cup C(j)$ , AFTER REVERSAL



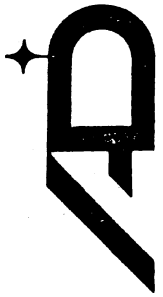
# INFLUENCE DIAGRAMS



REMOVAL OF CHANCE NODE  $j$  INTO VALUE NODE  $v$



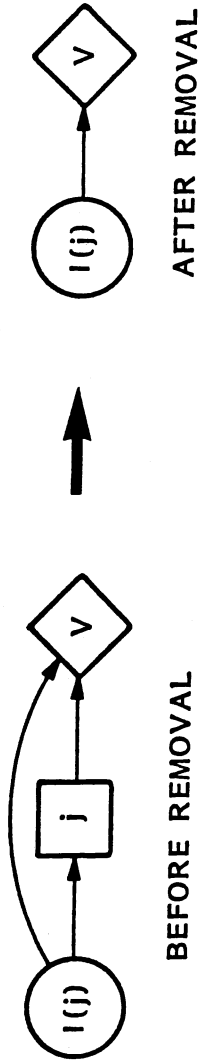
- EXPECTATION OF VALUE FUNCTION WITH RESPECT TO  $X_j$
- NODE  $j$  HAS SINGLE SUCCESSOR NODE  $v$



# INFLUENCE DIAGRAMS



## REMOVAL OF DECISION NODE $j$ INTO VALUE NODE $v$



- MAXIMIZATION OF VALUE FUNCTION BY CHOICE BY  $x_j$
- NODE  $j$  HAS SINGLE SUCCESSOR NODE  $v$
- $C(v) \subseteq \{j\} \cup i(j)$
- ALL NODES HAVING AN EFFECT ON VALUE NODE MUST BE OBSERVABLE





# MULTIVARIATE NORMAL DISTRIBUTION



## DEFINITIONS

$N = \{1, \dots, n\}$  , A SET OF INTEGERS

FOR  $j \in N$ ,  $X_j$  IS A SCALAR RANDOM VARIABLE

$X_N$  IS MULTIVARIATE NORMAL

- $E [X_N] = \mu_N$
- $\text{Cov} (X_N) = E [X_N X_N^T] - E [X_N] E [X_N^T] = \Sigma_{NN}$

CONDITIONAL DISTRIBUTIONS

- $\left\{ X_j \mid X_{C(j)} \right\}$  UNIVARIATE NORMAL
- $E \left[ X_j \mid X_{C(j)} = x_{C(j)} \right] = \mu_j + \Sigma_{jC(j)} \left( \Sigma_{C(j)C(j)} \right)^{-1} (x_{C(j)} - \mu_{C(j)})$
- $\text{VAR} \left[ X_j \mid X_{C(j)} = x_{C(j)} \right] = \Sigma_{jj} - \Sigma_{jC(j)} \left( \Sigma_{C(j)C(j)} \right)^{-1} \Sigma_{C(j)j}$



## NORMAL INFLUENCE DIAGRAM

### DEFINITIONS

- FOR CHANCE NODES  $j \in C$

$$E [X_j | X_{C(j)} = x_{C(j)}] = \mu_j + \sum_{k \in C(j)} b_{kj} (x_k - \mu_k)$$

$$\text{Var} [X_j | X_{C(j)} = x_{C(j)}] = v_j$$

$v_j = 0 \iff X_j$  DETERMINISTIC, LINEAR FUNCTION OF  $X_{C(j)}$

- FOR DECISION NODES  $j \in D$

$\mu_j$  = REFERENCE DECISION VALUE

$$v_j = 0$$

- VALUE NODE IS QUADRATIC

$$V(X_N) = 1/2 X_N^T Q X_N + P^T X_N + r$$



# NORMAL INFLUENCE DIAGRAM

## ASSESSMENT

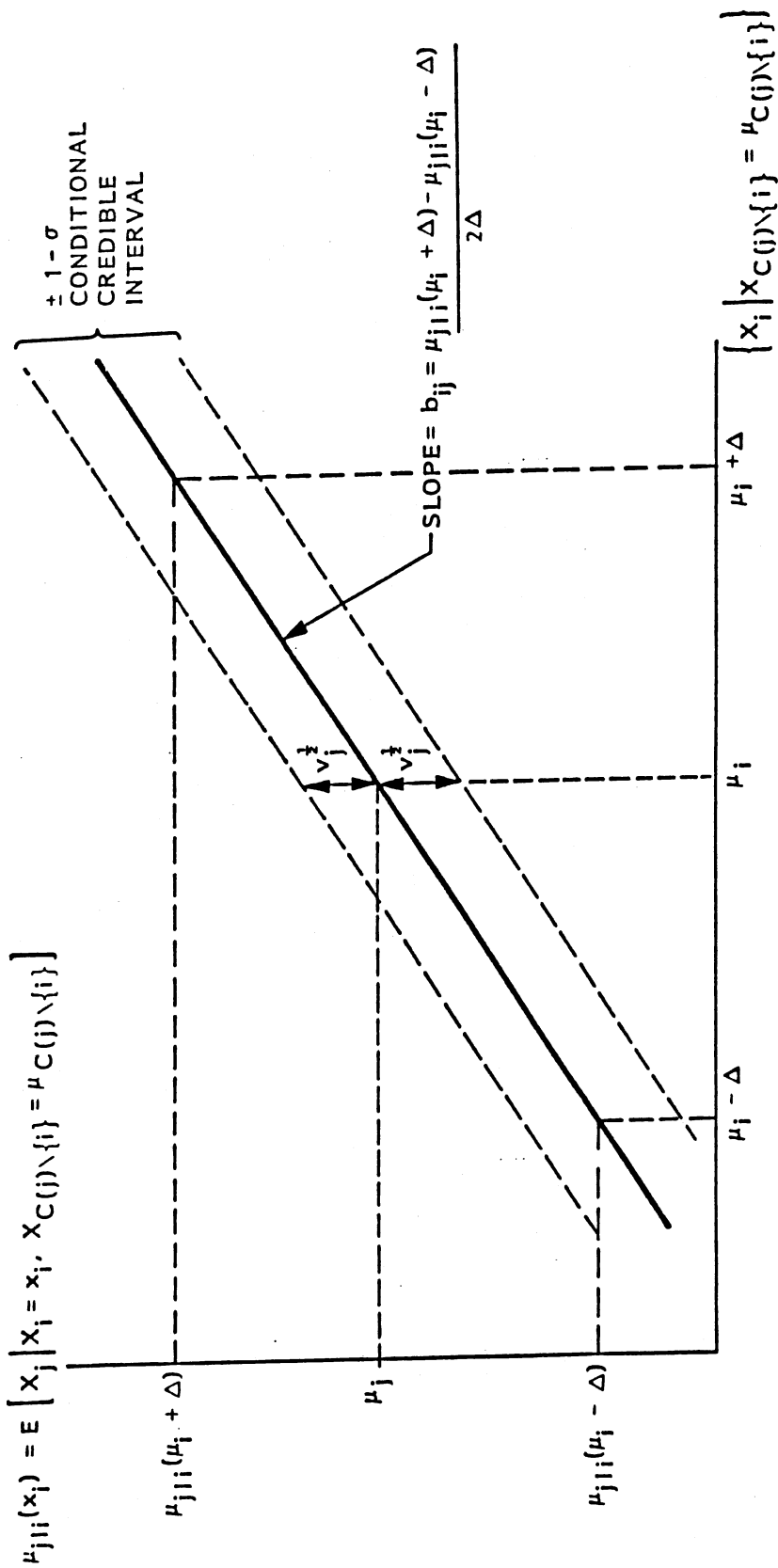
1. ASSESS AN INFLUENCE DIAGRAM  
DETERMINE VARIABLES  $X_N$   
DETERMINE GRAPH (C, D, C(j) FOR  $j \in C$ , I(j) FOR  $j \in D$ )
2. ORDER THE VARIABLES  $X_N$  SO THAT  
 $j \in C(k) \rightarrow j < k$   
 $j \in I(k) \rightarrow j < k$
3. FOR  $j = 1$  TO  $n$   
IF ( $j \in C$ )  
 $\mu_j = E [X_j | X_{C(j)} = \mu_{C(j)}]$   
 $\nu_j = \text{VAR} [X_j | X_{C(j)} = \mu_{C(j)}]$   
FOR  $i \in C(j)$   
 $b_{ij} = \frac{\partial E [X_j | X_{C(j) \setminus \{i\}} = \mu_{C(j) \setminus \{i\}} | X_i = x_i]}{\partial x_i}$   
END  
ELSE ( $j \in D$ )  
 $\mu_j =$  REFERENCE DECISION VALUE  
 $\nu_j = 0$   
ENDIF  
END



# NORMAL INFLUENCE DIAGRAM



## ASSESSMENT

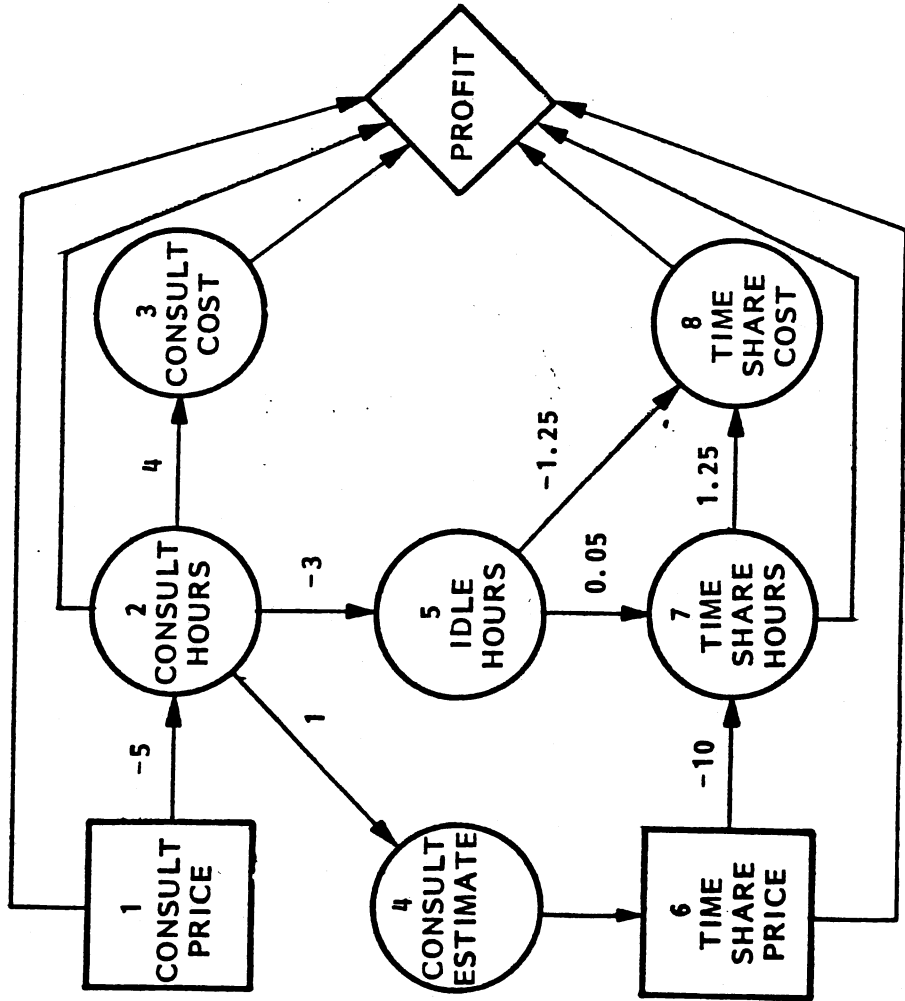




# NORMAL INFLUENCE DIAGRAM



## CONSULTANT'S PROBLEM



$$\text{PROFIT} = X_1 X_2 - X_3 + X_6 X_7 - X_8$$

$$N = \{1, 2, 3, 4, 5, 6, 7, 8\}$$

$$Q = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$p = (0 \ 0 \ -1 \ 0 \ 0 \ 0 \ 0 \ -1)^T$$

$$r = 0$$

□ = DECISION NODE

○ = CHANCE NODE

$$v = (0 \ 40,000 \ 4,000,000 \ 250,000 \ 100 \ 0 \ 10,000 \ 40,000)^T$$

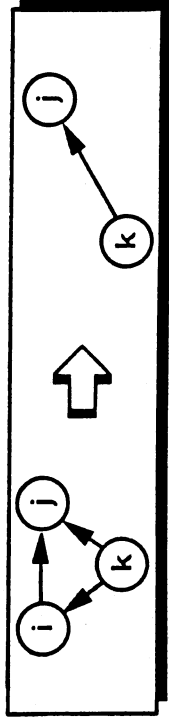
$$\mu = (100 \ 1,500 \ 58,000 \ 1,500 \ 3,500 \ 25 \ 750 \ 5,000)^T$$



## NORMAL INFLUENCE DIAGRAM



### REMOVAL OF NODE i INTO NODE j



- EXPECTATION WITH RESPECT TO NODE i
- NODE i MUST HAVE SINGLE DIRECT SUCCESSOR NODE j
- ALGORITHM

$$C(j) \leftarrow C(i) \cup C(j) \setminus \{i\}$$

$$b_{kj} \leftarrow b_{kj} + \frac{b_{ki} b_{ij}}{2} \text{ FOR } k \in C(i)$$

$$v_j \leftarrow v_j + b_{ij} v_i$$

$$N \leftarrow N \setminus \{i\}$$

- NOTE:  $v_j$  STAYS NON-NEGATIVE



# NORMAL INFLUENCE DIAGRAM



## REVERSAL OF ARC FROM NODE i TO NODE j

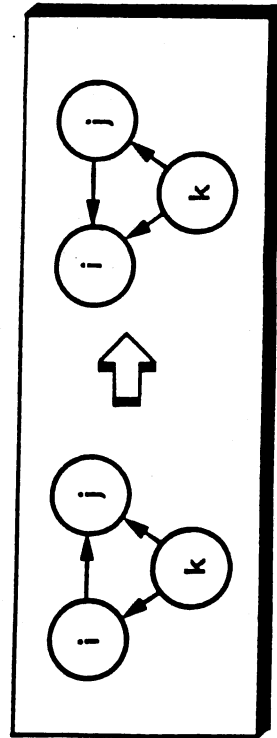
- INFLUENCE DIAGRAM VERSION OF BAYES' RULE
- $i \in C(j)$
- MUST HAVE NO OTHER DIRECTED PATH FROM i TO j (OTHERWISE ACYCLIC LOOP FORMED)
- ALGORITHM

$$C(j) \leftarrow C(i) \cup C(j) \setminus \{i\}$$

$$b_{kj} \leftarrow b_{kj} + b_{ki} b_{ij} \text{ for } k \in C(j)$$

$$v_j^{old} \leftarrow v_j$$

$$v_j \leftarrow v_j^{old} + b_{ij}^2 v_i$$



NOTE:  $v_i$  AND  $v_j$  STAY NON-NEGATIVE

IF  $v_j > 0$

$$C(i) \leftarrow C(j) \cup \{j\}$$

$$VRATIO \leftarrow v_i/v_j$$

$$v_i \leftarrow v_j^{old} \cdot VRATIO$$

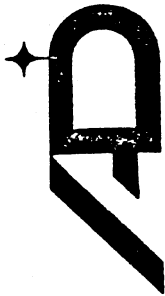
$$b_{ji} \leftarrow b_{ij} \cdot VRATIO$$

$$b_{kj} \leftarrow b_{ki} - b_{kj} b_{ji} \text{ for } k \in C(i)$$

ELSE

$$b_{ji} \leftarrow 0$$

END



# NORMAL INFLUENCE DIAGRAM



## REMOVING CHANGE NODE INTO VALUE NODE

- $d$  = SCALAR RANDOM VARIABLE WITH NO SUCCESSORS OBSERVABLE PRIOR TO REMAINING DECISIONS

- $s = N \setminus \{d\}$

- $Q = \begin{bmatrix} Q_{ss} & Q_{sd} \\ Q_{ds} & Q_{dd} \end{bmatrix}$

- ALGORITHM

$$Q_{ss} \leftarrow Q_{ss} + Q_{sd} B_{sd}^T + B_{sd} Q_{ds} + B_{sd} Q_{dd} B_{sd}^T$$

$$P_s \leftarrow P_s + B_{sd} P_d + (Q_{sd} + B_{sd} Q_{dd}) (\mu_d - B_{sd}^T \mu_s)$$

$$r \leftarrow r + \dagger Q_{dd} v_d + \dagger (\mu_d - B_{sd}^T \mu_s) Q_{dd} (\mu_d - B_{sd}^T \mu_s) + P_d (\mu_d - B_{sd}^T \mu_s)$$

$$N \leftarrow N \setminus \{d\}$$





## NORMAL INFLUENCE DIAGRAM



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### REMOVING DECISION NODE INTO VALUE NODE

- $X_d$  = CONTROLLABLE SCALAR VARIABLE SELECTED AFTER OBSERVING  $X_s$  WHEN  $s = N \setminus \{d\}$

- $Q_{dd}$  NEGATIVE SCALAR ELEMENT OF Q

- ALGORITHM

$$X_d^* (X_s) = - Q_{dd}^{-1} P_d - Q_{dd}^{-1} Q_{ds} X_s$$

$$Q_{ss} \leftarrow Q_{ss} - Q_{sd} Q_{dd}^{-1} Q_{ds}$$

$$P_s \leftarrow P_s - Q_{sd} Q_{dd}^{-1} P_d$$

$$r \leftarrow r - \frac{1}{2} P_d^T Q_{dd}^{-1} P_d$$

$$N \leftarrow N \setminus \{d\}$$



## NORMAL INFLUENCE DIAGRAM



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### SOLUTION PROCEDURE FOR CONSULTANT'S PROBLEM

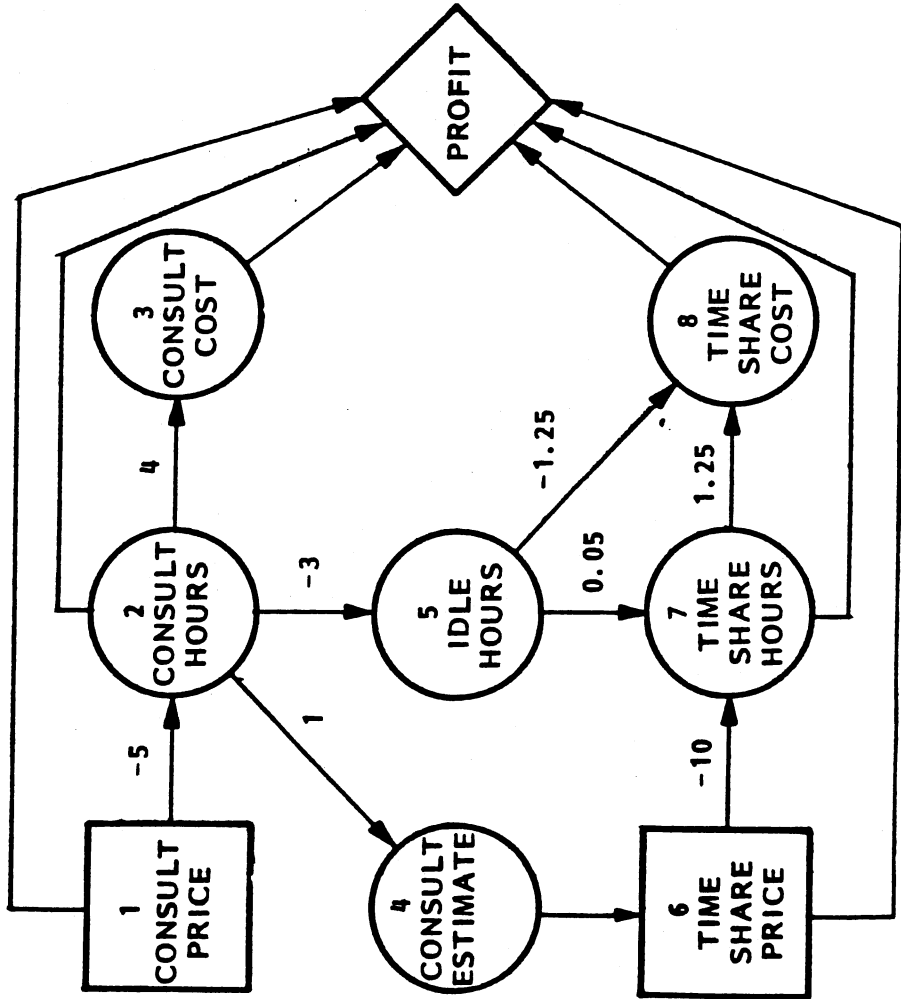
1. SUCCESSIVELY REMOVE CHANGE NODES  $x_8$ ,  $x_7$ ,  $x_5$ , AND  $x_3$  INTO THE VALUE NODE
2. REVERSE THE INFLUENCE FROM  $x_2$  TO  $x_4$
3. REMOVE CHANGE NODE  $x_2$  INTO THE VALUE NODE
4. REMOVE DECISION NODE  $x_6$  INTO THE VALUE NODE
5. REMOVE CHANGE NODE  $x_4$  INTO THE VALUE NODE
6. REMOVE DECISION NODE  $x_1$  INTO THE VALUE NODE



# NORMAL INFLUENCE DIAGRAM



## CONSULTANT'S PROBLEM



$$\text{PROFIT} = X_1 X_2 - X_3 + X_6 X_7 - X_8$$

$$N = \{1, 2, 3, 4, 5, 6, 7, 8\}$$

$$Q = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$p = (0 \ 0 \ -1 \ 0 \ 0 \ 0 \ 0 \ -1)^T$$

$$r = 0$$

□ = DECISION NODE

○ = CHANCE NODE

$$v = (0 \ 40,000 \ 4,000,000 \ 250,000 \ 100 \ 0 \ 10,000 \ 40,000)^T$$

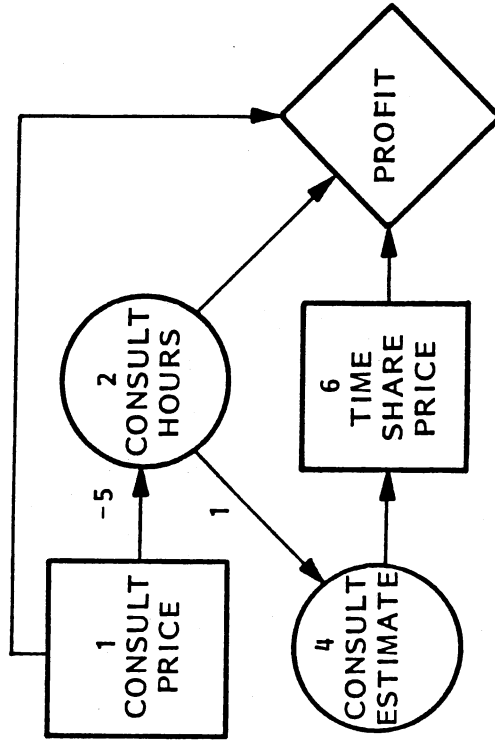
$$\mu = (100 \ 1,500 \ 58,000 \ 1,500 \ 3,500 \ 25 \ 750 \ 5,000)^T$$



# NORMAL INFLUENCE DIAGRAM

## REVERSAL STEP

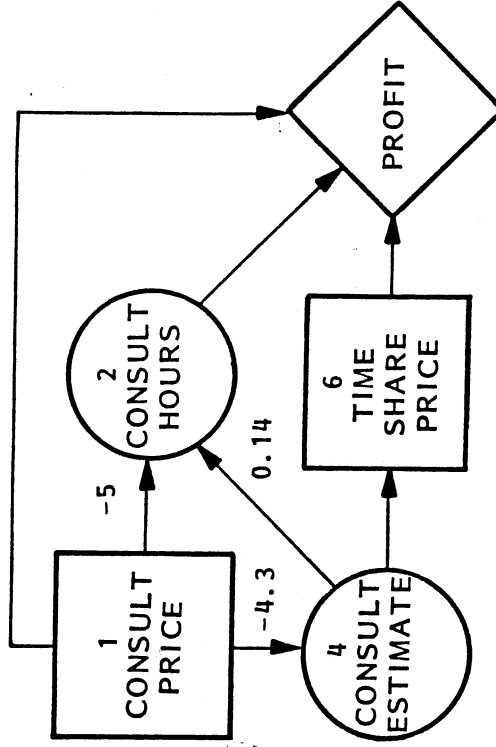
$N = \{1, 2, 4, 6, \}$



$V = (0 \ 40,000 \ 250,000 \ 0)^T$

BEFORE REVERSAL

$N = \{1, 4, 2, 6\}$



$V = (0 \ 290,000 \ 34,482.76 \ 0)^T$

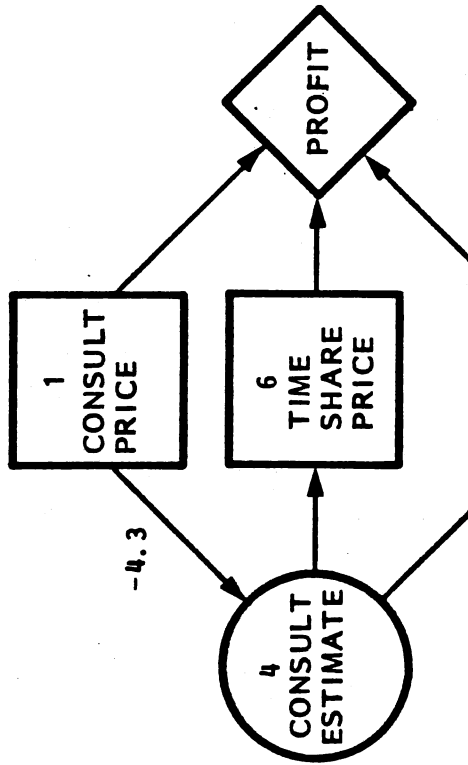
AFTER REVERSAL



# DISCRETE - TIME FILTERING



## MINIMAL REPRESENTATION OF CONSULTANT'S PROBLEM



$$N = \{1, 4, 6\}$$

$$Q = \begin{bmatrix} -8.6207 & 0.1379 & 0.6466 \\ 0.1379 & 0.0 & -0.0207 \\ 0.6466 & -0.0207 & -20.0 \end{bmatrix}$$

$$p = (1756.73 \ -1.043 \ 978.9)^T$$

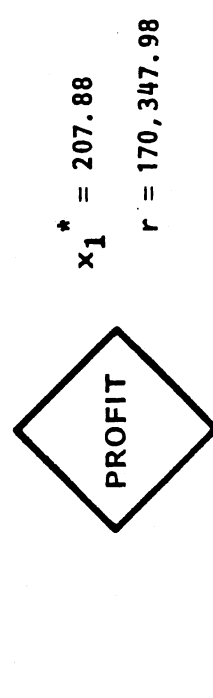
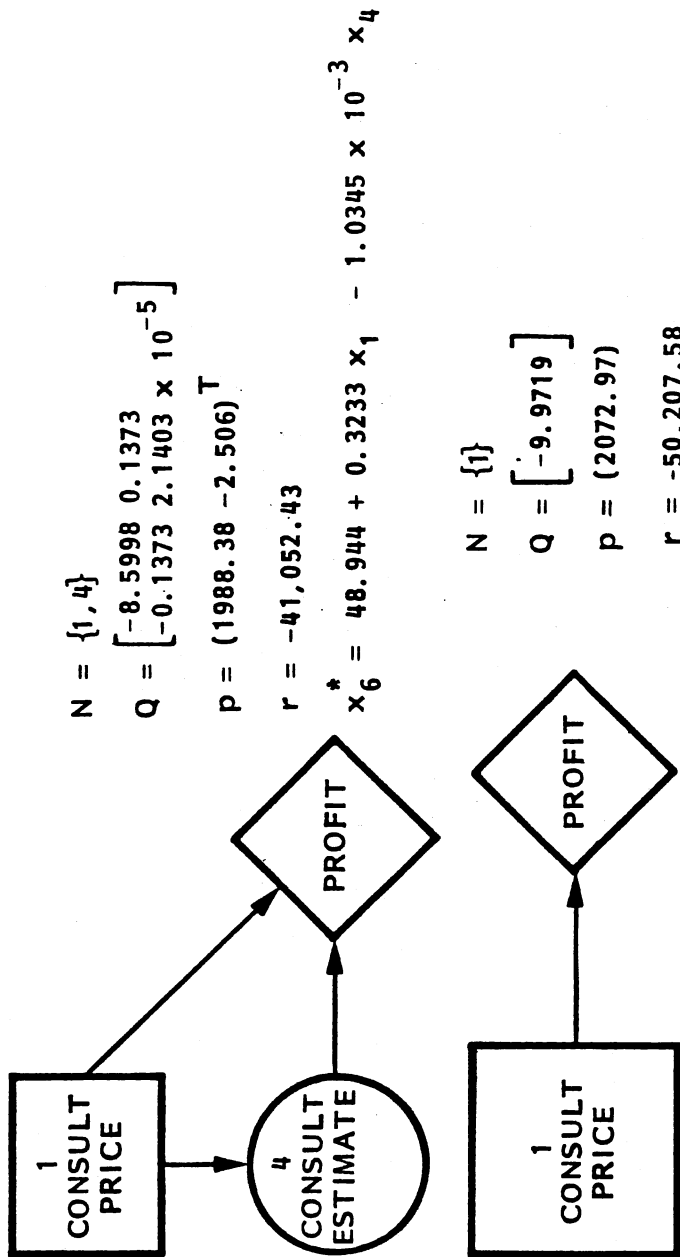
$$r = -65,007.54$$

$$v = (0 \ 34,482.76 \ 0)^T$$



# NORMAL INFLUENCE DIAGRAM

## OPTIMAL SOLUTION OF THE CONSULTANT'S PROBLEM

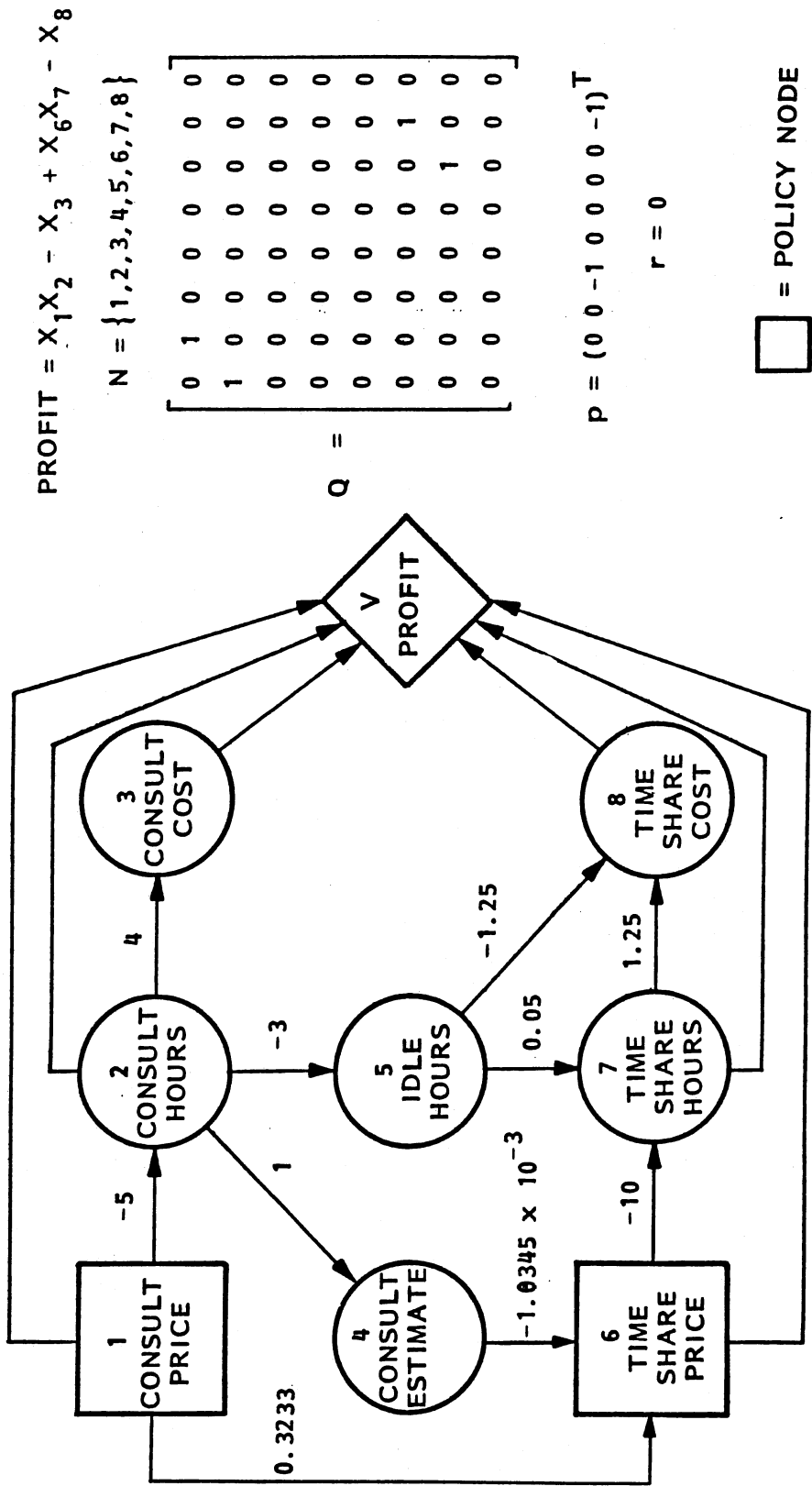




# NORMAL INFLUENCE DIAGRAM



## CONSULTANT'S POLICY DIAGRAM



$$\text{PROFIT} = X_1 X_2 - X_3 + X_6 X_7 - X_8$$

$$N = \{1, 2, 3, 4, 5, 6, 7, 8\}$$

$$Q = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$P = (0 \ 0 \ -1 \ 0 \ 0 \ 0 \ 0 \ -1)^T$$

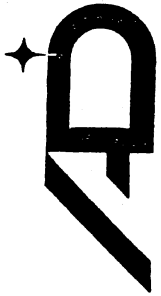
$$r = 0$$

□ = POLICY NODE

○ = CHANCE NODE

$$v = (0 \ 40,000 \ 4,000,000 \ 250,000 \ 100 \ 0 \ 10,000 \ 40,000)^T$$

$$\mu = (207.88 \ 960.59 \ 55842 \ 960.59 \ 5118.2 \ 54.67 \ 543.21 \ 2707.48)^T$$



# DISCRETE - TIME FILTERING



## MATHEMATICAL MODEL

DYNAMIC PROCESS:  $x(k+1) = \Phi(k)x(k) + \Gamma(k)w(k)$   $k = 0, \dots, N.$

MEASUREMENT PROCESS:  $z(k) = H(k)x(k) + v(k)$   $k = 0, \dots, N.$

PROBABILISTIC STRUCTURE:  $E[x(0)] = \mu_0.$   
 $\text{Cov}[x(0)] = P_0.$   
 $E[w(k)] = 0$  for  $k = 0, \dots, N.$

$\text{Cov}[w(j), w(k)] = \delta_{jk} Q_k$  for  $j = 0, \dots, N$  and  $k = 0, \dots, N.$   
 $Q_k$  are diagonal for  $k = 0, \dots, N.$

$\text{Cov}[x(0), w(0)] = 0.$

$E[v(k)] = 0$  for  $k = 0, \dots, N.$

$\text{Cov}[v(j), v(k)] = \delta_{jk} R_k$  for  $j = 0, \dots, N$  and  $k = 0, \dots, N.$   
 $R_k$  are diagonal for  $k = 0, \dots, N.$

$\text{Cov}[w(j), v(k)] = 0$  for  $j = 1, \dots, N$  and  $k = 0, \dots, N.$

$\text{Cov}[x(0), v(k)] = 0$  for  $0 = 1, \dots, N.$

DIMENSIONS OF VECTORS:  $x(k) \in R^n$ ,  $w(k) \in R^r$ ,  $z(k) \in R^p$ , and  $v(k) \in R^p.$

OBJECTIVE:  
FIND RECURSIVE ALGORITHM FOR  
GENERATING  $P\{x(k+1) | z(0), \dots, z(k)\}$   
FROM  $P\{x(k) | z(0), \dots, z(k-1)\}$  FOR  
 $k = 0, \dots, N$

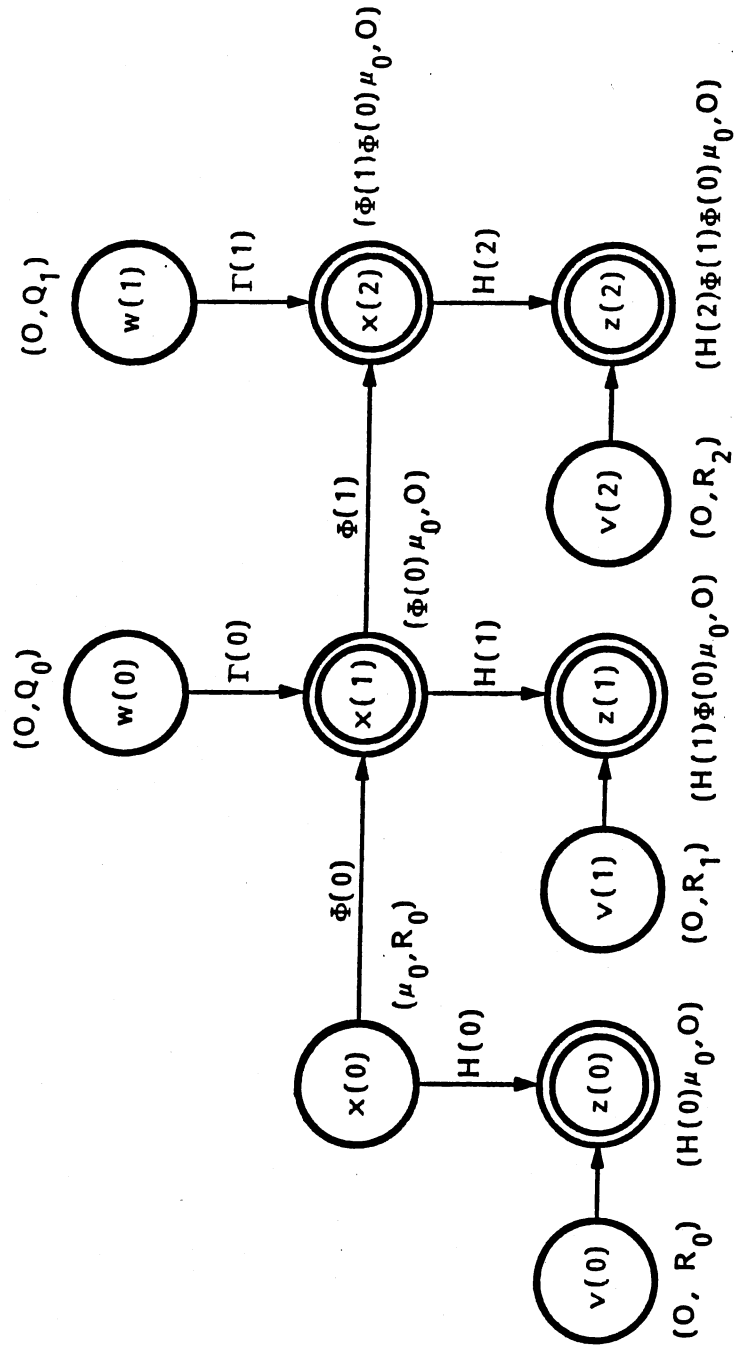




# DISCRETE - TIME FILTERING



## INFLUENCE DIAGRAM REPRESENTATION



○ = VECTOR CHANGE NODE

⊙ = VECTOR DETERMINISTIC NODE



## DISCRETE - TIME FILTERING



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### INCORPORATION OF MEASUREMENT $z(k)$

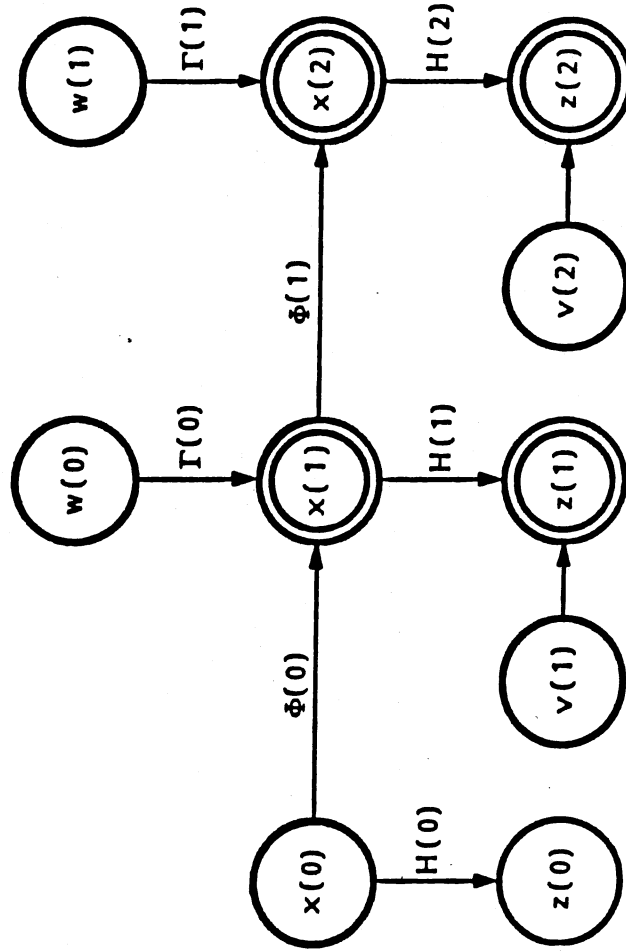
- A. MEASUREMENT UPDATE
  - 1. REMOVE  $v(k)$  INTO  $z(k)$
  - 2. REVERSE ARCS FROM  $x(k)$  TO  $z(k)$
  - 3. INSTANTIATE  $z(k)$  AND UPDATE  $E[x(k)]$
- B. TIME UPDATE
  - 1. REMOVE  $x(k)$  INTO  $x(k+1)$
  - 2. REMOVE  $w(k)$  INTO  $x(k+1)$



# DISCRETE - TIME FILTERING



$V(0)$  REMOVED INTO  $Z(0)$



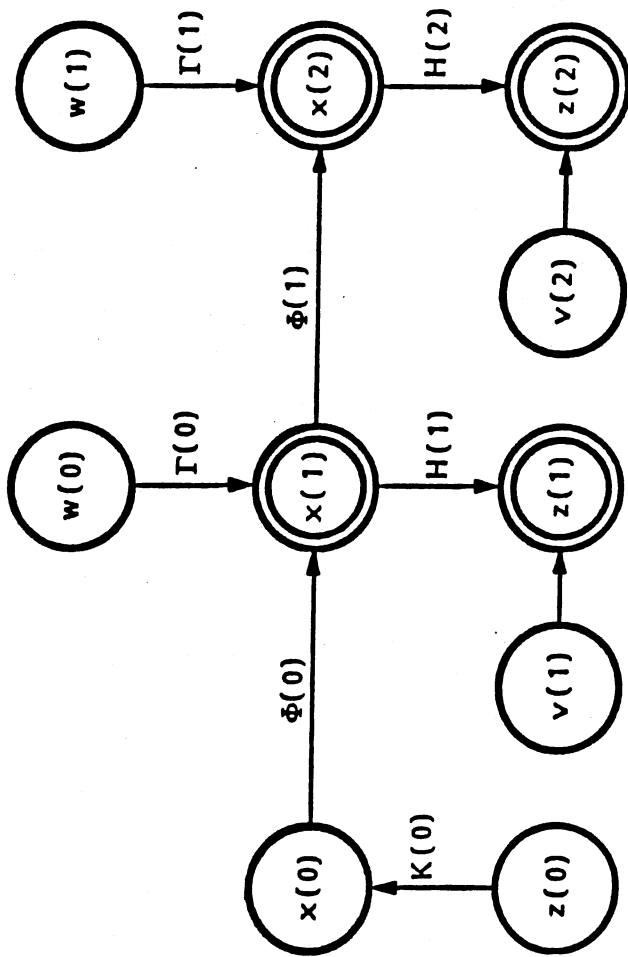
- = VECTOR CHANCE NODE
- = VECTOR DETERMINISTIC NODE



# DISCRETE - TIME FILTERING



X(0) AND Z(0) REVERSED



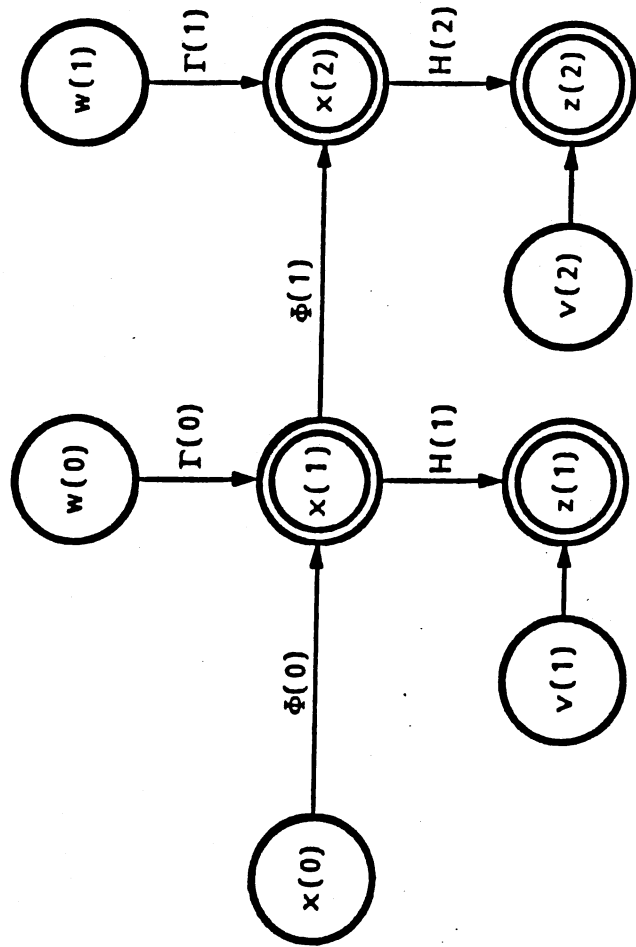
- = VECTOR CHANGE NODE
- ⊙ = VECTOR DETERMINISTIC NODE



# DISCRETE - TIME FILTERING



Z(0) INSTANTIATED



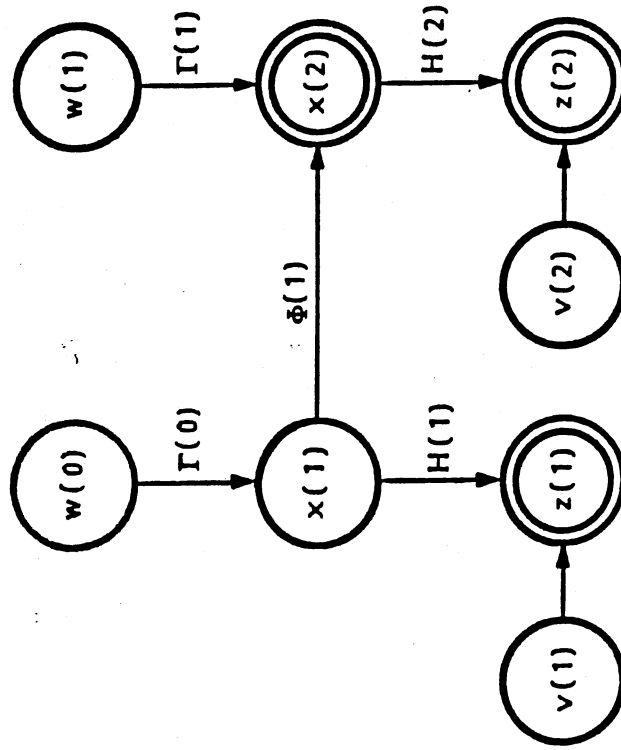
-  = VECTOR CHANGE NODE
-  = VECTOR DETERMINISTIC NODE



# DISCRETE - TIME FILTERING



X(0) REMOVED INTO X(1)

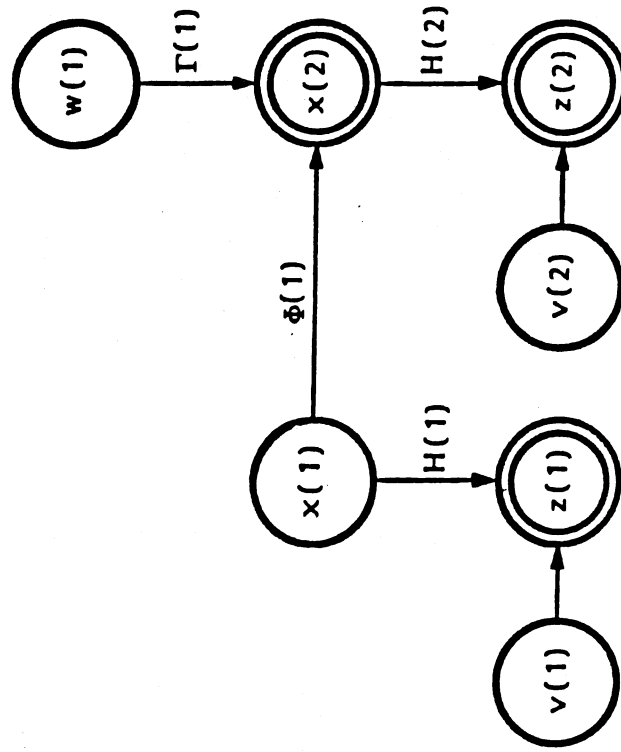


-  = VECTOR CHANGE NODE
-  = VECTOR DETERMINISTIC NODE



# DISCRETE - TIME FILTERING

W(0) REMOVED INTO X(1)



- $\circ$  = VECTOR CHANGE NODE
- $\odot$  = VECTOR DETERMINISTIC NODE



# DISCRETE - TIME FILTERING



## OPERATION COUNTS FOR PROCESSING A VECTOR OF p MEASUREMENTS

|                                    | ALGORITHM              | ADDITIONS                                  | MULTIPLICATIONS                            | DIVISIONS | SQUARE ROOTS |
|------------------------------------|------------------------|--|--|-----------|--------------|
| SCALAR<br>PROCESSING<br>ALGORITHMS | INFLUENCE DIAGRAM      | $(1.5n^2 + 1.5n)p$                         | $(1.5n^2 + 4.5n)p$                         | $np$      | 0            |
|                                    | CONVENTIONAL KALMAN    | $(1.5n^2 + 1.5n)p$                         | $(1.5n^2 + 4.5n)p$                         | p         | 0            |
|                                    | U-D COVARIANCE         | $(1.5n^2 + 3.5n)p$                         | $(1.5n^2 + 5.5n)p$                         | $np$      | 0            |
|                                    | SQUARE ROOT COVARIANCE | $(1.5n^2 + 3.5n)p$                         | $(2n^2 + 5n)p$                             | $2np$     | $np$         |
|                                    | POTTER SQUARE ROOT     | $(3n^2 + 3n)p$                             | $(3n^2 + 4n)p$                             | $2p$      | p            |
|                                    | KALMAN STABILIZED      | $(4.5n^2 + 5.5n)p$                         | $(4n^2 + 7.5n)p$                           | p         | 0            |
| BATCH<br>PROCESSING<br>ALGORITHMS  | SRIF, R TRIANGULAR     | $(n^2 + 2n)p + 1.5n^2 + 2.5n$              | $(n^2 + 3n)p + 2n^2 + 3n$                  | 2n        | n            |
|                                    | NORMAL EQUATION        | $(0.5n^2 + 1.5n)p + (n^3 + 6n^2 + 5n) / 6$ | $(0.5n^2 + 2.5n)p + (n^3 + 9n^2 - 4n) / 6$ | 2n        | n            |
|                                    | SRIF, R GENERAL        | $n^2p + (4n^3 + 3n^2 + 5n) / 6$            | $n^2p + (4n^3 + 9n^2 - n) / 6$             | 2n        | n            |





# DISCRETE - TIME FILTERING



## WEIGHTED OPERATION COUNTS FOR PROCESSING A VECTOR OF p MEASUREMENTS

| ALGORITHM              | WEIGHTED OPERATION COUNTS                    |
|------------------------|--|
| INFLUENCE DIAGRAM      | $(3.6n^2 + 12.3n)p$                          |
| CONVENTIONAL KALMAN    | $(3.6n^2 + 7.8n + 4.5)p$                     |
| U-D COVARIANCE         | $(3.6n^2 + 15.7n)p$                          |
| SQUARE ROOT COVARIANCE | $(4.3n^2 + 40.9n)p$                          |
| POTTER SQUARE ROOT     | $(7.2n^2 + 8.6n + 30.4)p$                    |
| KALMAN STABILIZED      | $(10.1n^2 + 16n + 4.5)p$                     |
| SRIF, R TRIANGULAR     | $(2.4n^2 + 6.2n)p + 4.3n^2 + 37.1n$          |
| NORMAL EQUATION        | $(1.2n^2 + 5.0n)p + 0.4n^3 + 3.1n^2 + 30.3n$ |
| SRIF, R GENERAL        | $2.4n^2p + 1.6n^3 + 2.6n^2 + 31n$            |

| OPERATION | WEIGHT |
|-----------|--------|
| +         | 1      |
| X         | 1.4    |
| ÷         | 4.5    |
| √         | 21.4   |



## DISCRETE - TIME FILTERING

### OPERATION COUNTS FOR TIME UPDATE

| ALGORITHM              | ADDITIONS  | MULTIPLICATIONS   | DIVISIONS                  | SQUARE ROOTS |
|------------------------|--|---|----------------------------|--------------|
| INFLUENCE DIAGRAM      | $1.17n^3 + 0.17n^2 - 3n + 3 + (2.5n^2 - 2.5n + 1)r + (n - 1)r^2$ | $1.17n^3 + 1.3n^2 - 4.5n + 5 + (2.5n^2 + 0.5n - 1)r + (n - 1)r^2$ | $0.5n^2 + 0.5n + (n - 1)r$ | 0            |
| CONVENTIONAL KALMAN    | $1.5n^3 + 2n^2 + 0.5n + (0.5n^2 + 0.5n)r$                        | $1.5n^3 + 1.5n^2 + (0.5n^2 + 1.5n)r$                              | 0                          | 0            |
| U-D COVARIANCE         | $1.5n^3 + 0.5n^2 + n^2r$   | $1.5n^3 + 2.5n^2 - n + (n^2 + 2n - 1)r$                           | $n - 1$                    | 0            |
| SQUARE ROOT COVARIANCE | $1.7n^3 + 2n^2 + 0.3n + (n^2 + n)r$                              | $1.7n^3 + 2n^2 + 0.3n + (n^2 + n)r$                               | $n$                        | $n$          |



# DISCRETE - TIME FILTERING



## WEIGHTED OPERATION COUNTS FOR TIME UPDATE

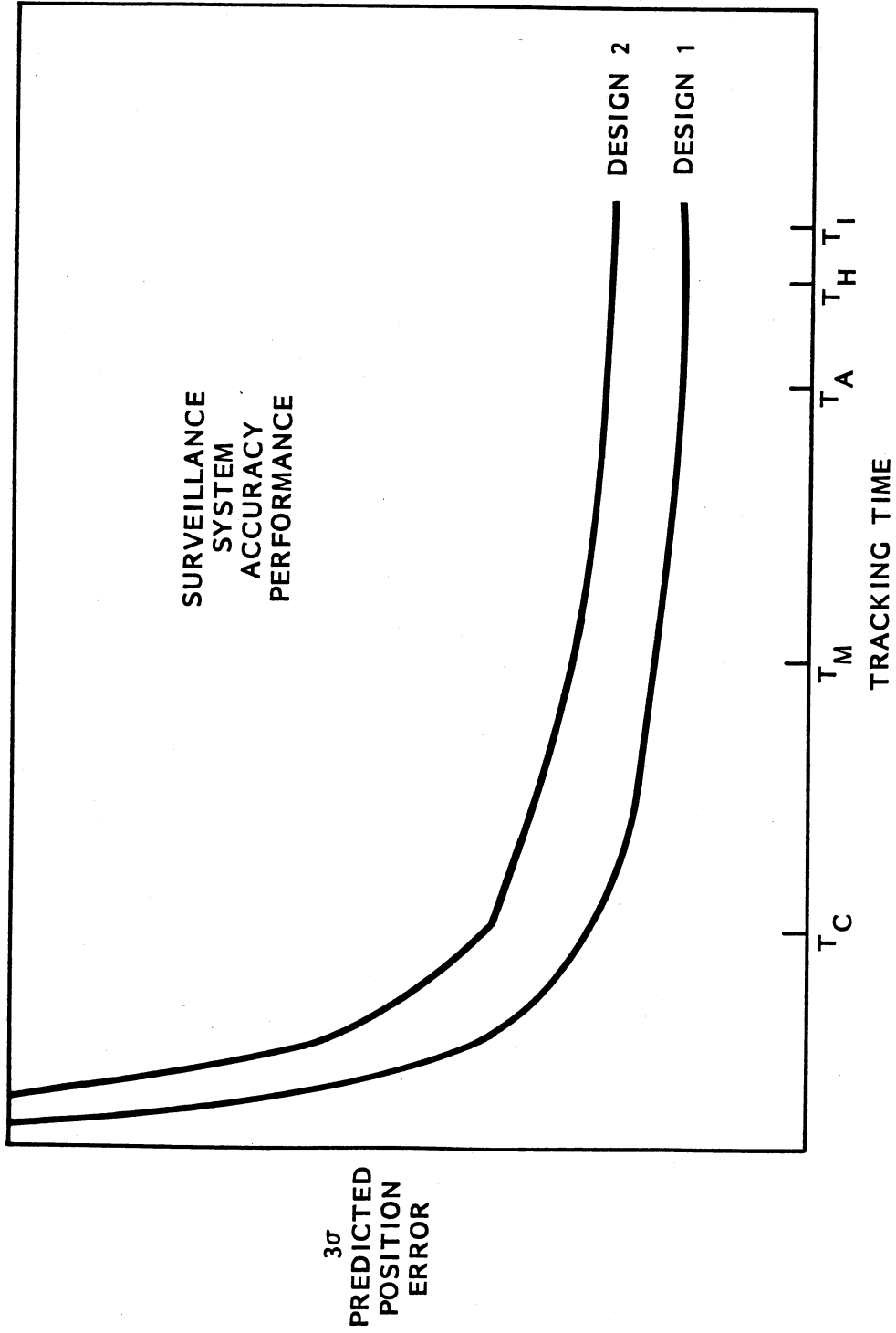
| ALGORITHM              | WEIGHTED OPERATION COUNTS   |
|------------------------|---|
| INFLUENCE DIAGRAM      | $2.8n^3 + 3.95n^2 - 11.55n + 10 + (6n^2 + 2.7n - 4.9)r + (2.4n - 2.4)r^2$ |
| CONVENTIONAL KALMAN    | $3.6n^3 + 4.1n^2 + 0.5n + (1.2n^2 + 2.6n)r$                               |
| U-D COVARIANCE         | $3.6n^3 + 4n^2 + 3.1n - 4.5 + (2.4n^2 + 4.2n - 2.8)r$                     |
| SQUARE ROOT COVARIANCE | $4n^3 + 4.8n^2 + 26.7n + (2.4n^2 + 2.4n)r$                                |



# $\Delta V$ -ACCURACY ANALYSIS



TYPICAL ACCURACY ANALYSIS OUTPUT

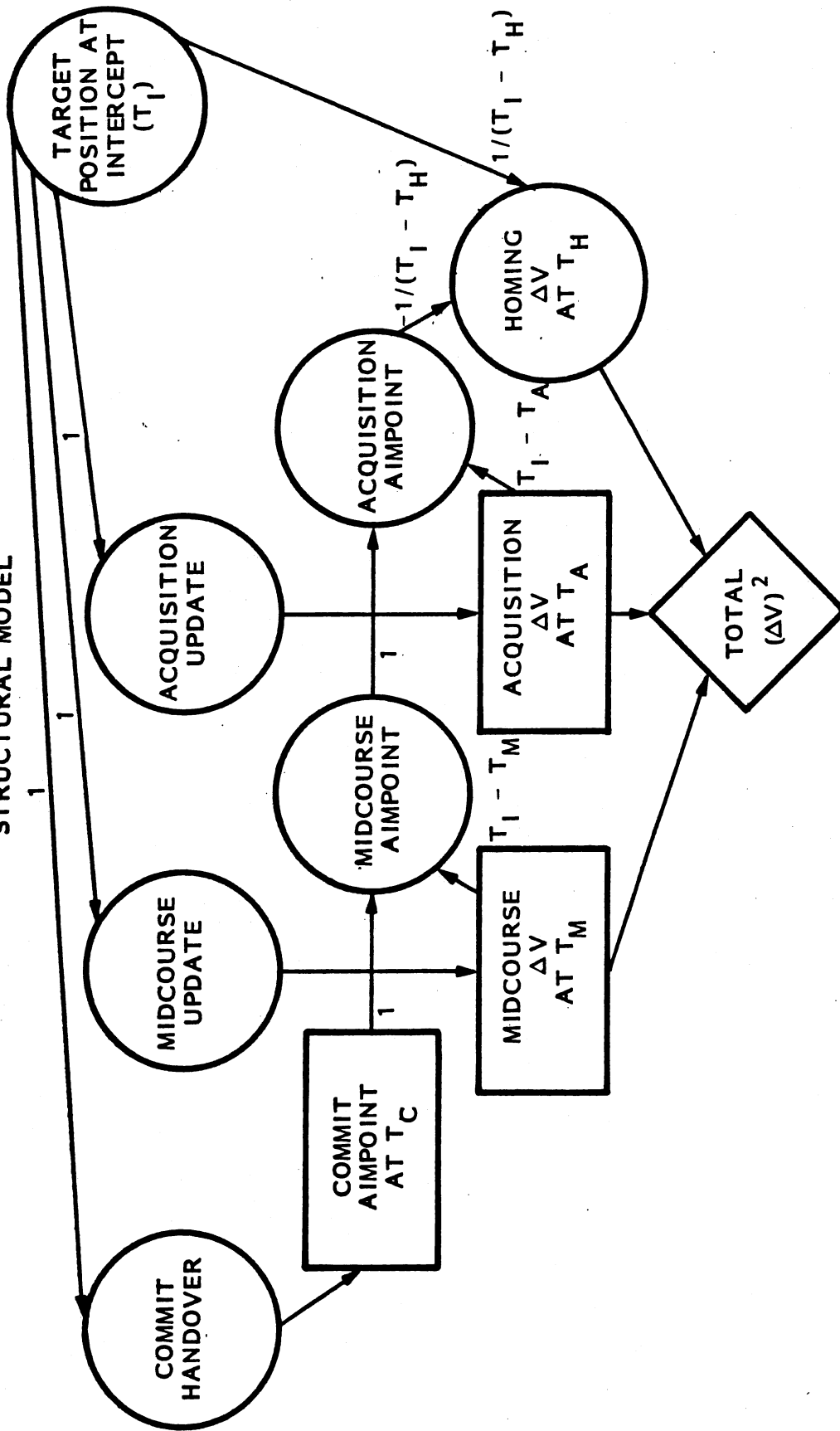




# $\Delta V$ -ACCURACY ANALYSIS



STRUCTURAL MODEL





# $\Delta V$ -ACCURACY ANALYSIS

