



World-Wide Crude Transportation Logistics :

A Decision Support System based on Simulation and Optimization

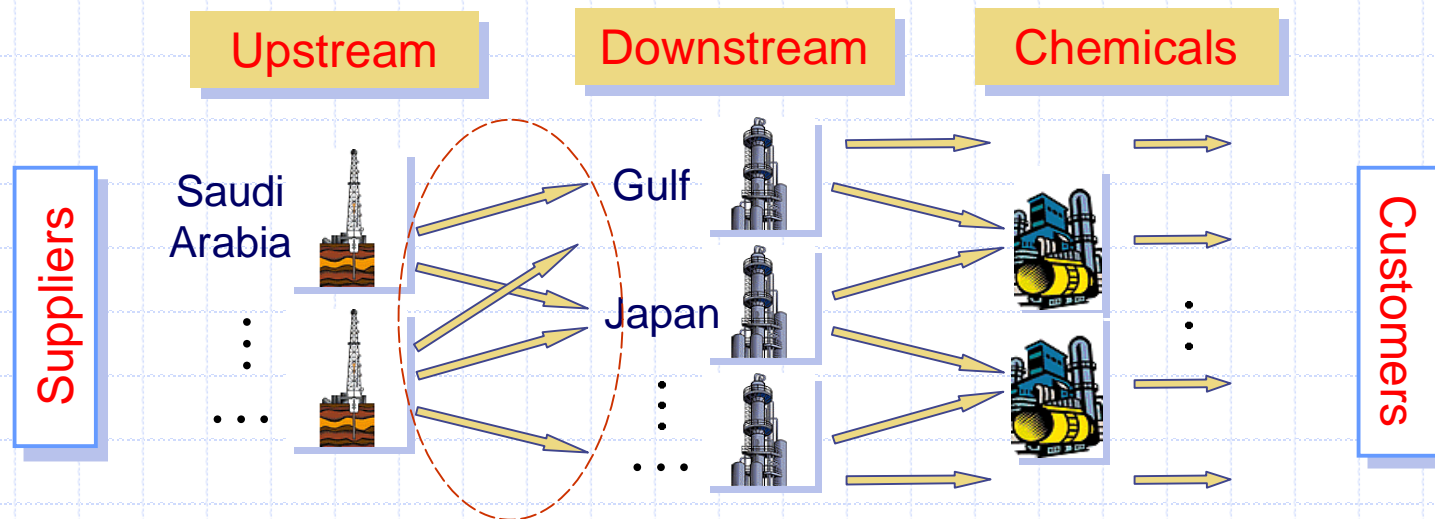
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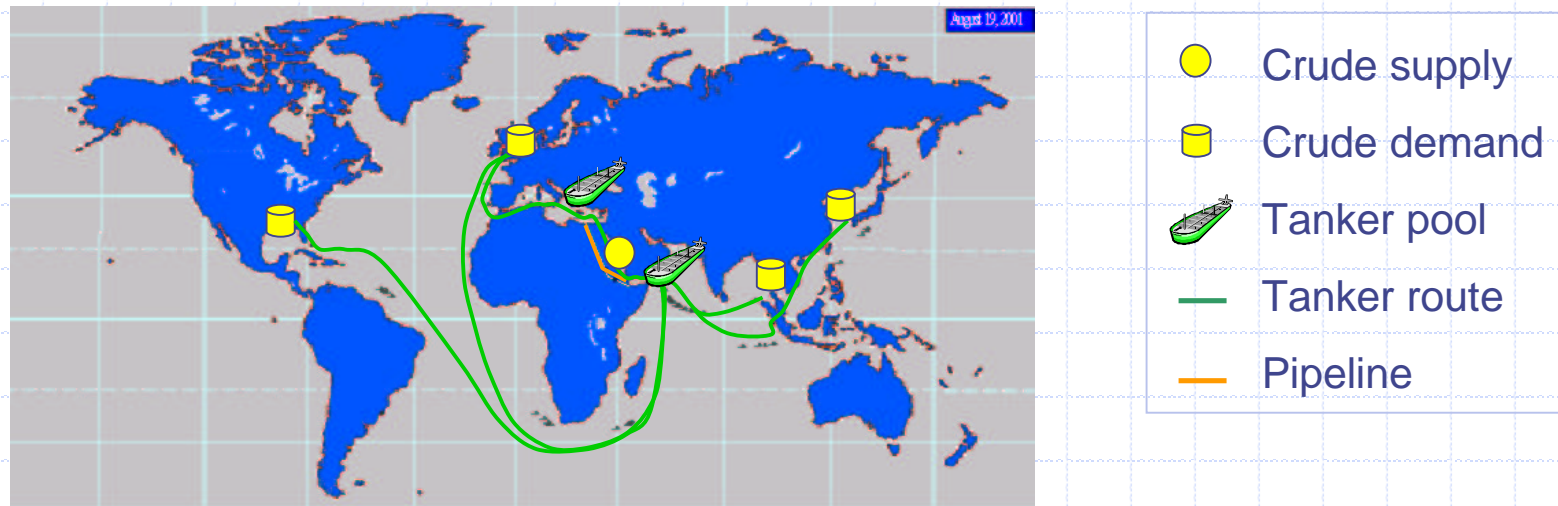
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Motivation and Objectives

- Transportation is the central operation between the “upstream” and “downstream” functions
- To investigate the behavior and improve the performance of the combined transportation and inventory system through simulation and optimization



World Wide Crude Logistics



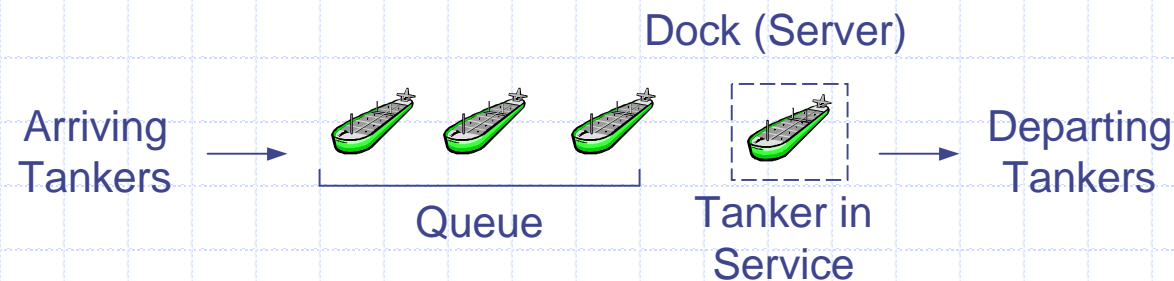
Assume:

- Yearly world-wide crude transportation cost \sim 2 billion \$
- Crude transportation through tanker fleet and pipeline
- Decisions: Sizing and composition of the tanker fleet; dispatch and routing of tankers

Methodologies

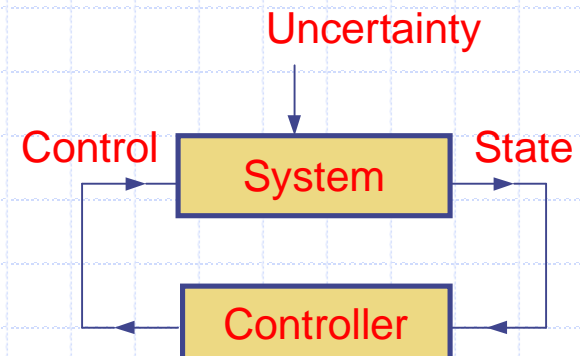
- Discrete event simulation

→ Conduct experiments with the model of a real system to investigate the system behavior and evaluate operation strategies



- Stochastic optimal control

→ Decision makers periodically observe the state of the system and choose control actions according to certain policies

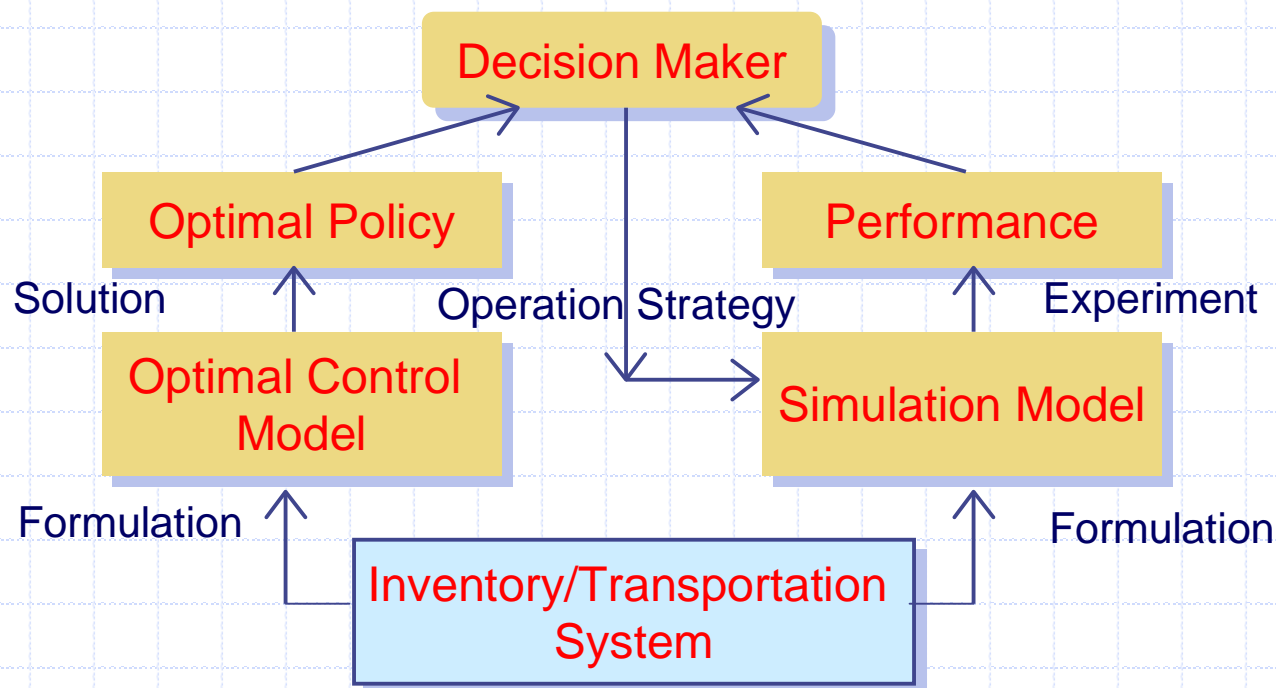




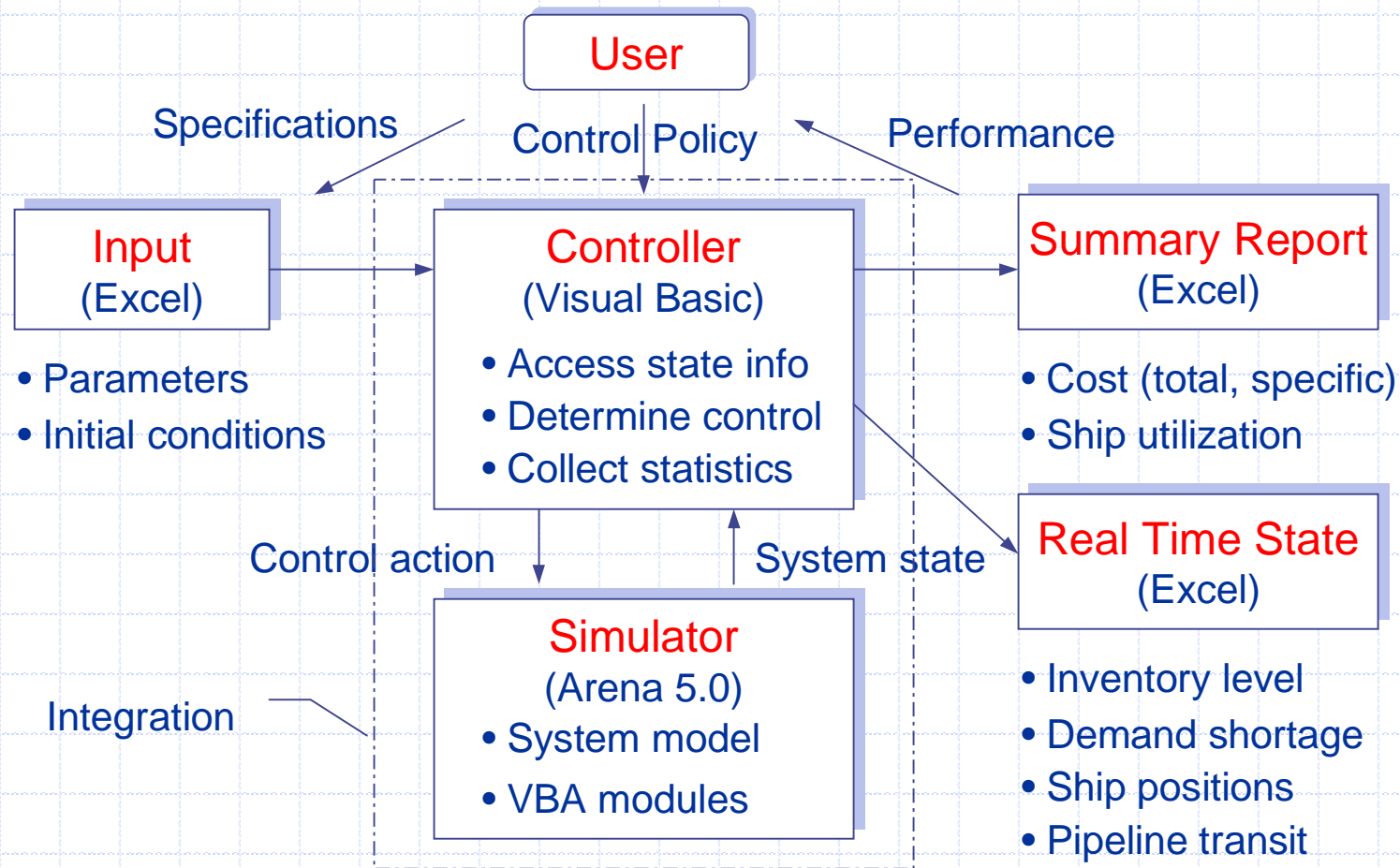
Discrete Event Simulation

Decision Support System

- An integrated decision support system
 - Simulation: Investigate behavior and evaluate performance
 - Optimization: Find the optimal or near optimal policies

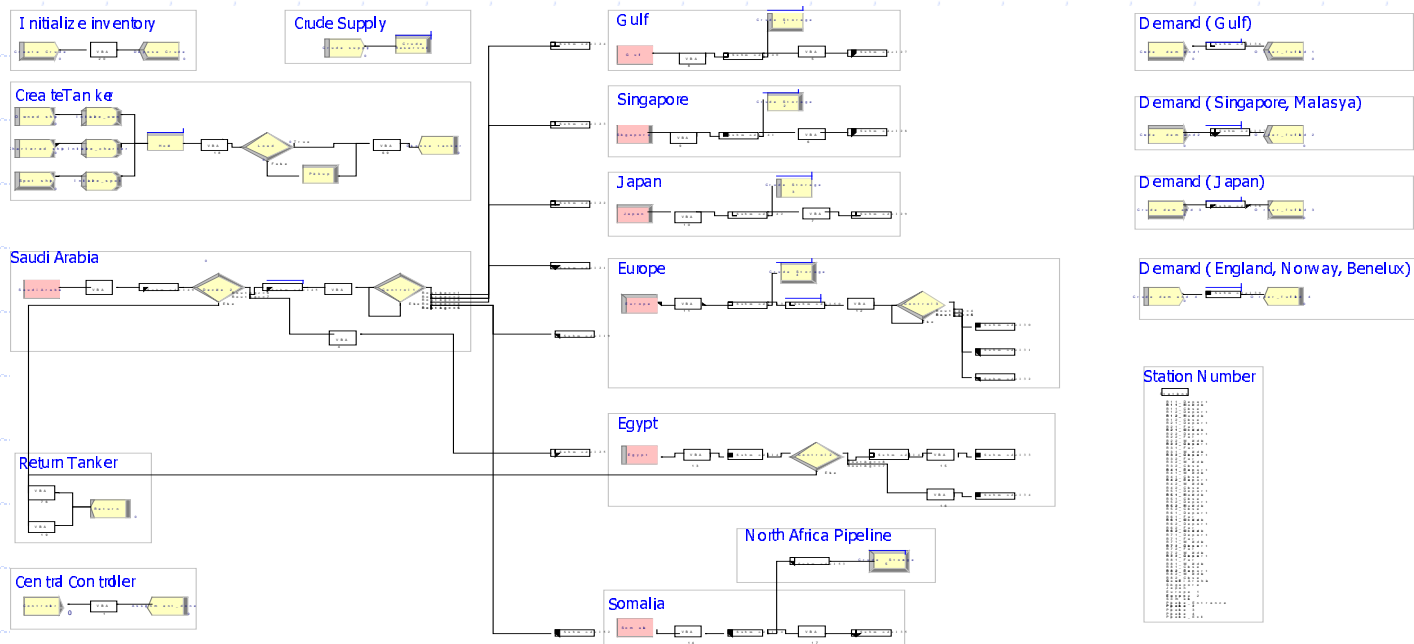


Simulation - Framework



Simulation - Model

- Discrete event simulation using Arena®
 - Define the model using a process oriented approach
 - Perform the simulation using an event driven approach



Controller Design

- Integration of simulation model and controller
 - Access and monitor state information in simulation
 - Manipulate variables or perform actions
- Active X automation through Visual Basic for Application
- Design and operate the tanker fleet

Physical actions	Arena actions
Rent/return tankers	Create/dispose entities
Dispatch tankers	Send release signals
Route tankers	Assign "route" attribute

Simulation - Animation

- Visualize the entity flow through in the dynamic system
 - Movement: traveling tankers, crude parcel, etc
 - Queuing: crude inventory, waiting orders, etc



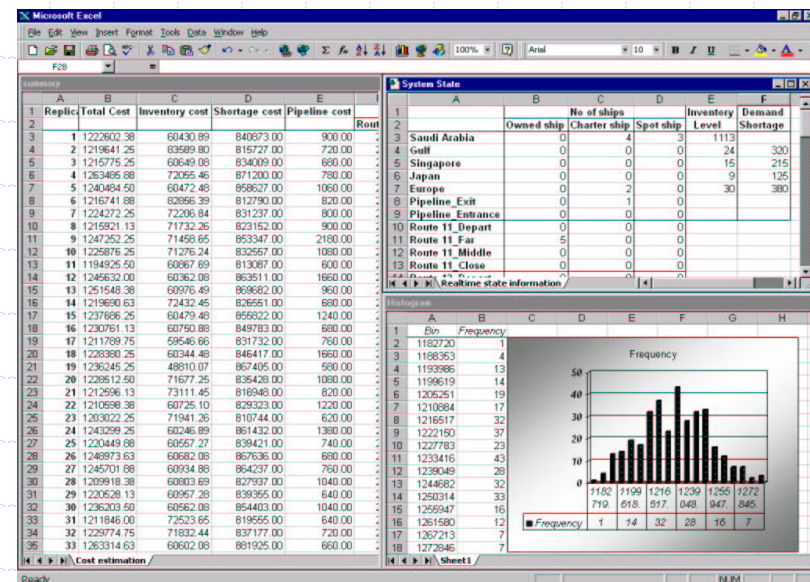
Simulation - Report

■ Summary report

- Cost information - total cost, cost per route, cost per type of tanker, cost per barrel, etc.
- System statistics - tanker utilization, etc.
- Stochastic analysis

■ Real-time state report

- Inventory levels
- Demand shortage
- Positions of tankers
- Status of pipeline



Statistical Analysis

- Analyze and interpret simulation results
- Evaluate system performance under various design and operation strategies
- Example, "How many spot tanker shall we rent?"

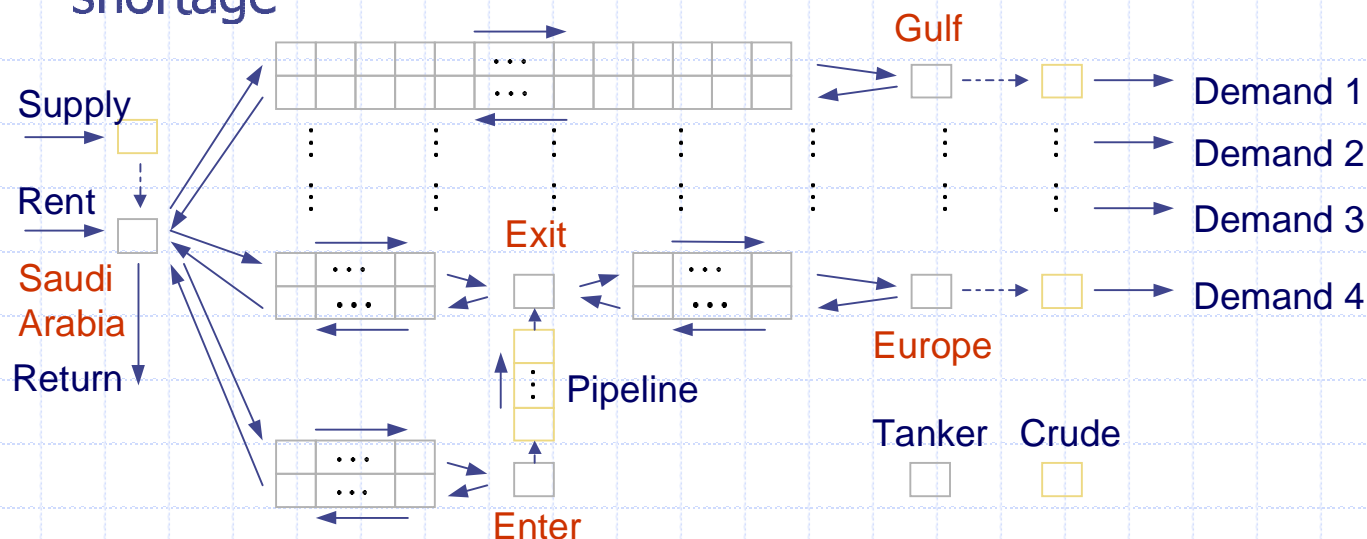
	Number of spot tanker	Total cost	Shortage penalty	Ship Utilization
1	0	784656.50	122665.95	59%
2	10	775240.81	8603.27	60%
3	20	841588.52	8578.35	52%



Stochastic Optimal Control

Transportation-Inventory System

- Inventory routing problem
 - A central decision maker is responsible for replenishing inventory by managing a fleet
 - Coordination between inventory control and vehicle routing
 - Transportation system serves as a buffer against demand shortage

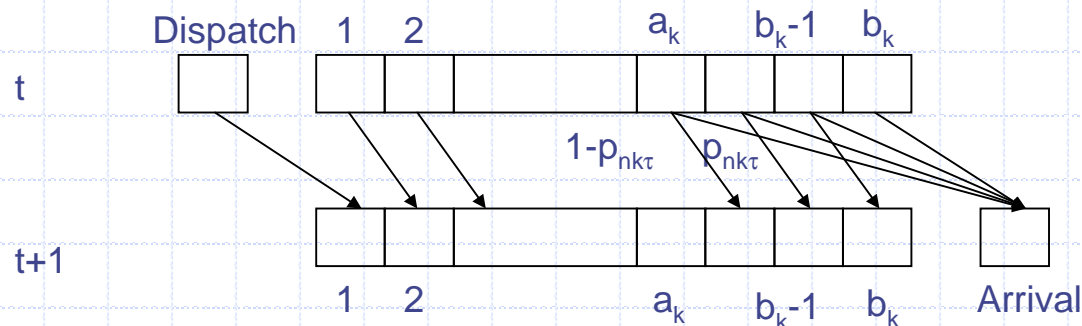


Uncertainty in Travel Time

- Incorporate uncertainty in tanker travel time into inventory/transportation system

$$\alpha_{nk\tau} = \begin{cases} 1 & \text{if tankers that have traveled on route } (n, k) \text{ for } \tau \text{ days} \\ & \text{arrive today given that they have not arrived before} \\ 0 & \text{otherwise} \end{cases}$$

$$p_{nk\tau} = P\{\alpha_{nk\tau} = 1\} = \frac{F_{nk}(\tau+1) - F_{nk}(\tau)}{F_{nk}(\infty) - F_{nk}(\tau)} \quad F_{nk}(t) \quad \text{Probability distribution of travel time}$$



Stochastic Optimal Control

γ State information

→ Inventory levels, Demand shortages, Tanker positions, Status of pipeline transportation

γ Control decisions

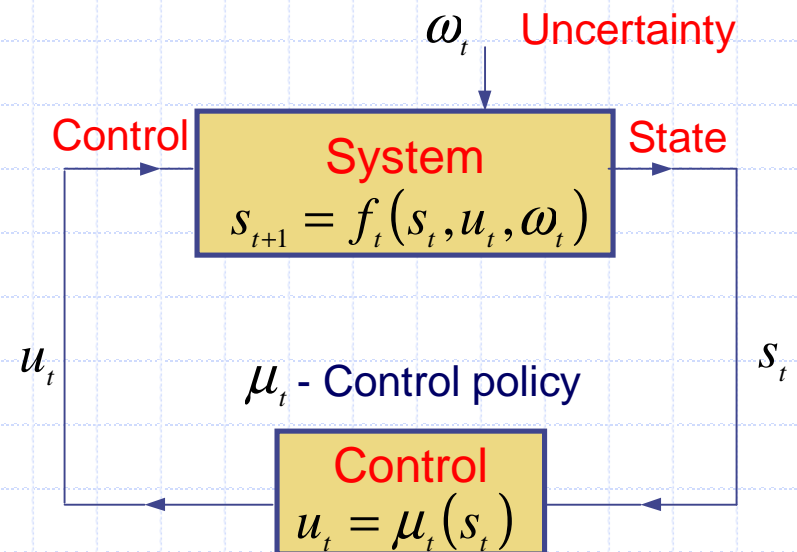
→ Dispatch and routing of existing fleet

→ Creating (chartering or spotting) and returning of tankers

γ Uncertainties

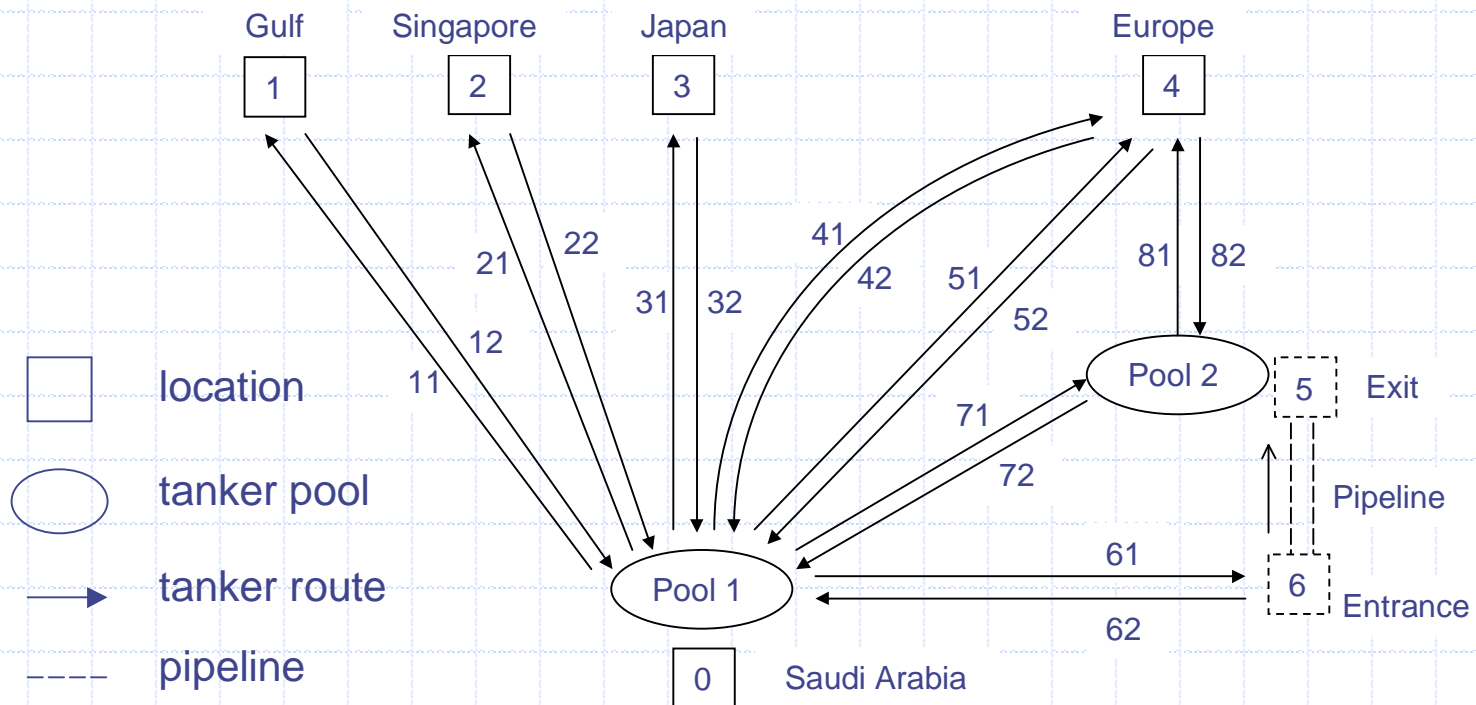
→ Crude demands, Tanker travel time, etc.

Minimize: $E \{ \text{Transportation cost} + \text{Inventory cost} + \text{Shortage penalty} \dots \}$



Schematic Model

- A simplified representation of the combined inventory/transportation system



Mathematical Formulation

- System state $s(t)$

- $w_{ij}(t)$ number of tankers j at location i
- $x_{nkj\tau}(t)$ number of tankers j traveling on route (n,k) for τ days
- $z_i(t)$ inventory level at location i
-

- Control action $u(t)$

- $y_{nkj}(t)$ number of tankers j dispatched to route (n,k)
- $r_{ij}(t)$ number of tankers j rented/returned at location i

- Random variable $\omega(t)$

- $d_i(t)$ number of tankers j rented/returned at location i
- $\alpha_{nk\tau}(t)$ =1 if tankers j that have been traveled on route (n,k) for τ days arrive

Mathematical Formulation (cont.)

- State equations $s(t+1) = f_t(s(t), u(t), \omega(t))$

$$w_{0j}(t+1) = w_{0j}(t) - \sum_{k=1}^7 y_{1kj}(t) + \sum_{k=1}^7 \sum_{\tau=a_k}^{b_k} \alpha_{2k\tau} x_{2kj\tau}(t) + r_{0j}(t) \quad \text{tanker pool}$$

$$z_0(t+1) = z_0(t) - \sum_{j=1}^3 \sum_{k=1}^6 y_{1kj}(t) \cdot c + v(t) \quad \text{inventory change}$$

... ..

- Cost equations $g_t(s(t), u(t))$

tanker costs + operating costs + holding costs + inventory costs + shortage penalty + canal toll + pipeline costs ...

- Control Set $u(t) \in U(s(t))$

$$0 \leq \sum_{k=1}^7 y_{1kj}(t) \leq w_{0j}(t) + r_{0j}(t) \quad \text{tanker availability}$$

$$\sum_{j=1}^3 y_{21j}(t) \leq (MC_1 - z_1(t)) / c \quad \text{storage availability}$$

Dynamic Programming

- Optimal control is to find a strategy π , which is a sequence of control policies

$$\pi = (\mu_1, \dots, \mu_N) \in \Pi \quad u_t = \mu_t(s_t)$$

such that

$$J_{\pi^*}(s_0) = \min_{\pi \in \Pi} E \left\{ \sum_{t=1}^{N-1} g_t(s_t, \mu_t(s_t), \omega_t) + g_N(s_N) \right\}$$

- Bellmans' equations

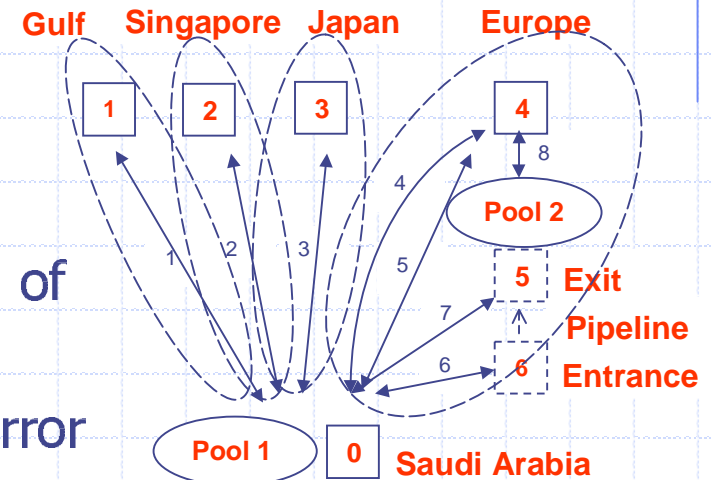
$$J_t(s_t) = \min_{u_t \in U(s_t)} \{ g_t(s_t, u_t) + E\{J_{t+1}(s_{t+1}) | s_t, u_t\} \}$$

$$\mu_t(s_t) = \arg \min_{u_t \in U(s_t)} \{ g_t(s_t, u_t) + E\{J_{t+1}(s_{t+1}) | s_t, u_t\} \}$$

$\forall s_t \in S$ $N(s)$ is prohibitively large – “curse of dimensionality”

Solution Strategies

- Decomposition of the system
 - A controller for each subsystem
- Approximation of subproblems
 - Parametric function approximation of individual cost function
 - Find a parameter that minimizes error



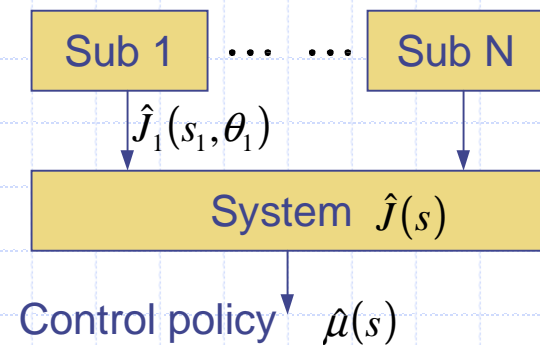
$$\tilde{J}_{mt}(s_{mt}, v_{mt}, \theta_t) = \theta_{1t} \phi_1(s_{mt}, v_{mt}) + \dots + \theta_{Kt} \phi_K(s_{mt}, v_{mt})$$

v_{mt} tanker capacity assigned to subsystem m θ_t parameter vector

$$\min_{\theta_t} \sum_{(s_{mt}, v_{mt}) \in S} \gamma_{mt}(s_{mt}, v_{mt}) [J_{mt}(s_{mt}, v_{mt}) - \tilde{J}_{mt}(s_{mt}, v_{mt}, \theta_t)]^2$$

Solution Strategies (cont.)

- Computation of control policies
 - Approximate the overall cost function by solving a knapsack problem
 - Find near optimal policies based on the Bellman equations

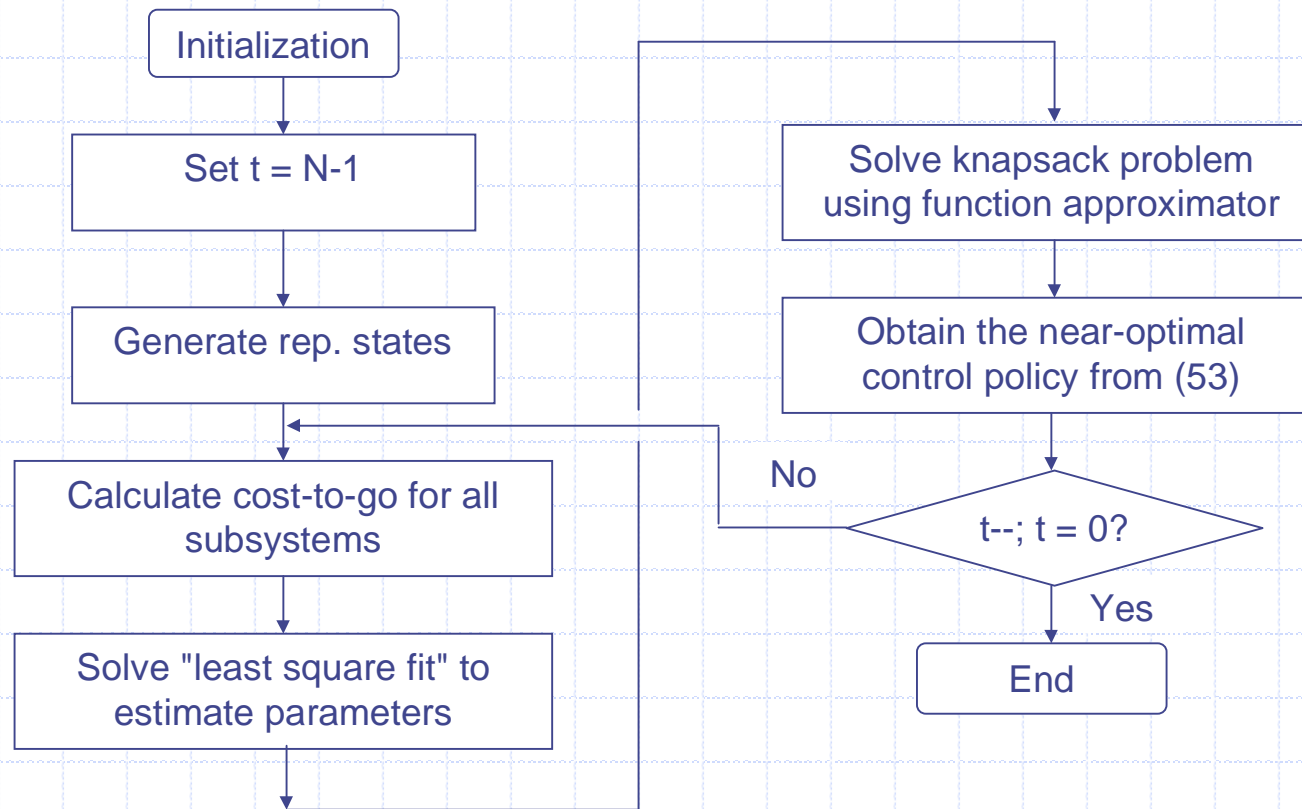


$$\hat{J}_t(s_t) = \min_{v_{mt} \in Z_+^1} \left\{ \sum_{m=1}^4 J_{mt}(s_{mt}, v_{mt}) + r_{02} \cdot CC \right\}$$

$$\text{s.t.} \quad \sum_{m=1}^4 v_{mt} \leq \begin{pmatrix} w_{01t} \\ w_{02t} + r_{02} \\ w_{03t} \end{pmatrix}$$

$$\hat{\mu}_t(s_t) = \arg \min_{u_t \in U(s_t)} \left\{ g_t(s_t, u_t) + E \left\{ \hat{J}_{t+1}(s_{t+1}) \mid s_t, u_t \right\} \right\}$$

Approximate Algorithm





Suboptimal Control Schemes

Refinery Operation



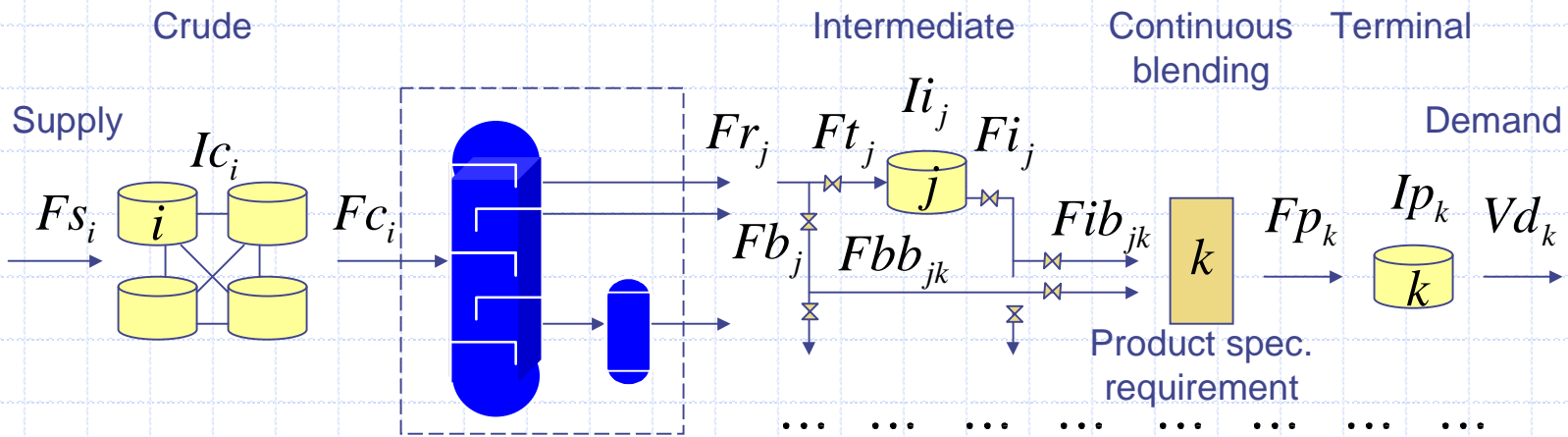
Assume: A typical large refinery

- ~ 3-10 Crude tanks
- ~ 10-30 Intermediate tanks
- ~ 30-50 Product tanks
- ~ 2 week demand amount of inventory

Assume:

- Tank investment ~ 3 million \$/each; Tank yearly operating cost ~ 150 thousand \$/each
- Continuous blending technology makes it possible to produce finished products only when they are needed
- Decisions: ordering/delivery, real-time plant operation

Model Formulation



- ◆ Inventory $I(t)$: Ic_i - crude; Ii_j - intermediate; Ip_k - product
- ◆ Flowrates $F(t)$: Fc_i - feed; Ft_j, Fb_j - tank/blender splits;
 Fbb_{jk}, Fib_{jk} - blend recipes
- ◆ Equations: e.g. $Ip(k, t) = Ip(k, t - 1) - Vd(k, t - 1) + Fp(k, t - Lt - 1) \dots$
 Lt - transportation lead time
- ◆ Objective: $Cost = \alpha \sum_t I(t) \bar{P}(t) + \sum_t S(t) \bar{C}(t) \dots$ α - interest rate

Moving Horizon Control (sub-optimal)

- State information

- Inventory levels (crude, intermediate, finished)
- Crude supply/Product demand
- Production yields

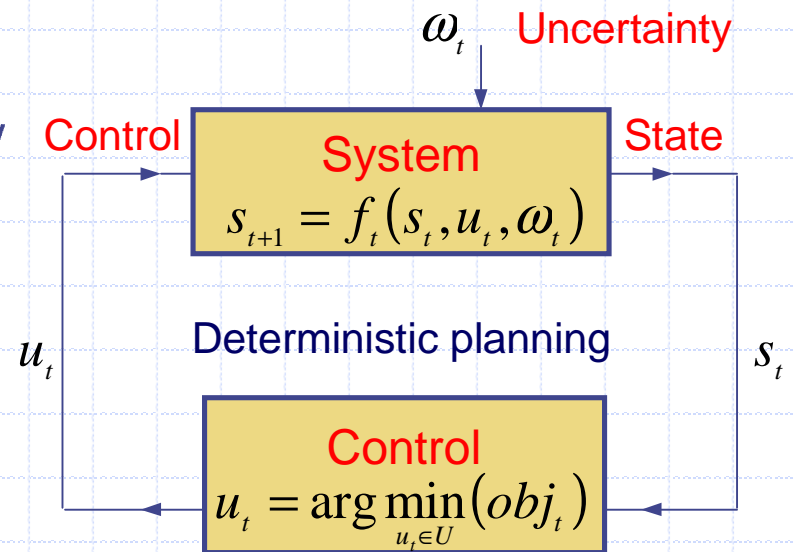
- Control decisions

- Purchase/Delivery
- Plant operation (feed, tank/blend splits, blend recipes)

- Uncertainties

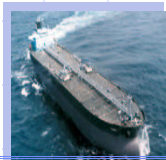
- Crude supply availability
- Production yields
- Product stream prices

Minimize: Inventory/Tankage cost
+ Shortage penalty



Conclusions

- Systematic decision makings in the supply chain management through simulation and optimization



Supply Chain Optimization

Crude logistics system

- ◆ Coordination between transportation and inventory control
 - ◆ Simulation based investigation and evaluation
 - ◆ Optimal control of the transportation/inventory system
 - ◆ Approximation architecture for computing policies
- Currently working on model verification validation and efficiency improvement
 - Could be implemented in a world wide crude logistics system