



Dynamic Stochastic Model for a Single Airport Ground Holding Problem

Avijit Mukherjee
Mark Hansen

University of California at Berkeley



Literature

- ❑ Static Integer Programming
 - ❑ Many papers on deterministic problem
 - ❑ Stochastic problem studied in Richetta and Odoni (1993); Hoffman (1997); Ball et al.(2003)
 - ❑ Equity issues addressed by Vossen et al. (2002)



Literature

- ☐ Dynamic Models: Richetta and Odoni (1994)
 - ☐ Inability to revise previously assigned ground delays
 - ☐ Objective function is to minimize expected delay cost
 - ☐ No longer Linear Programming Problem if non-linear measure of delay introduced
 - ☐ Can handle specific type of scenario tree



Research Contributions

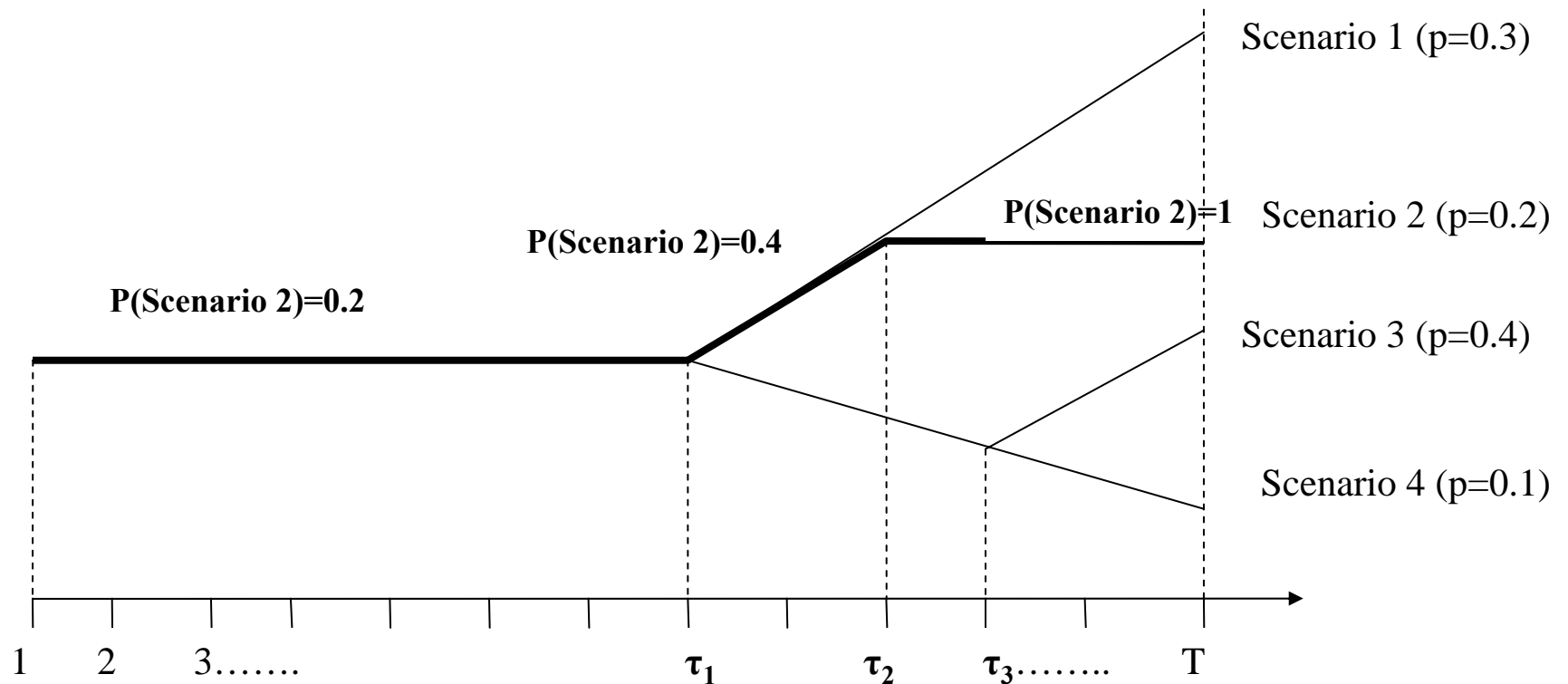
- ☐ **Dynamic Stochastic Model for SAGHP**
 - ☐ Ability to revise ground delays of some flights (non-departed)
 - ☐ Can handle any generalized scenario tree

- ☐ **Alternative Objective Functions**
 - ☐ Expected Squared Deviation from RBS Allocation

- ☐ **Multi-Criteria Optimization**

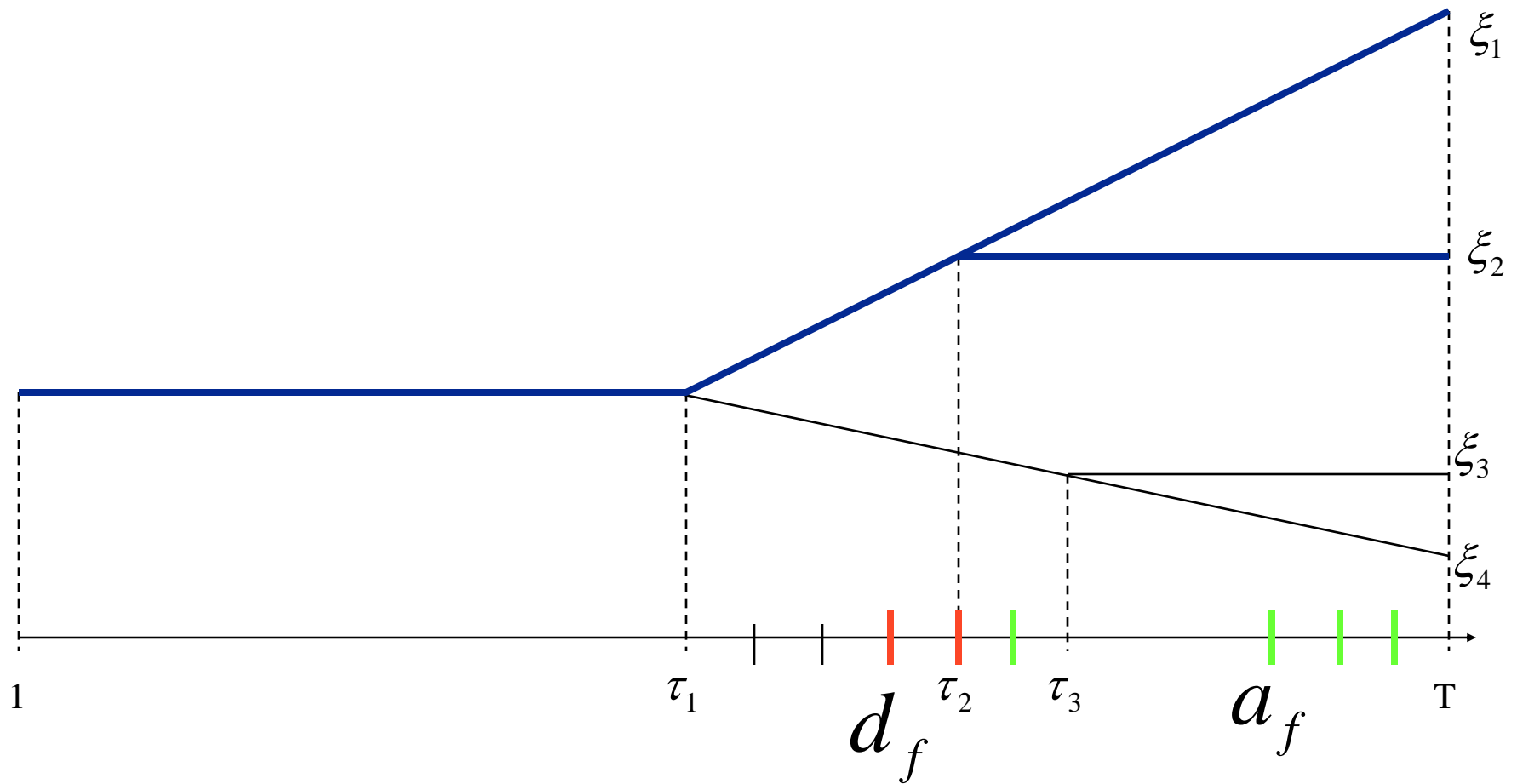


Capacity Scenario Tree





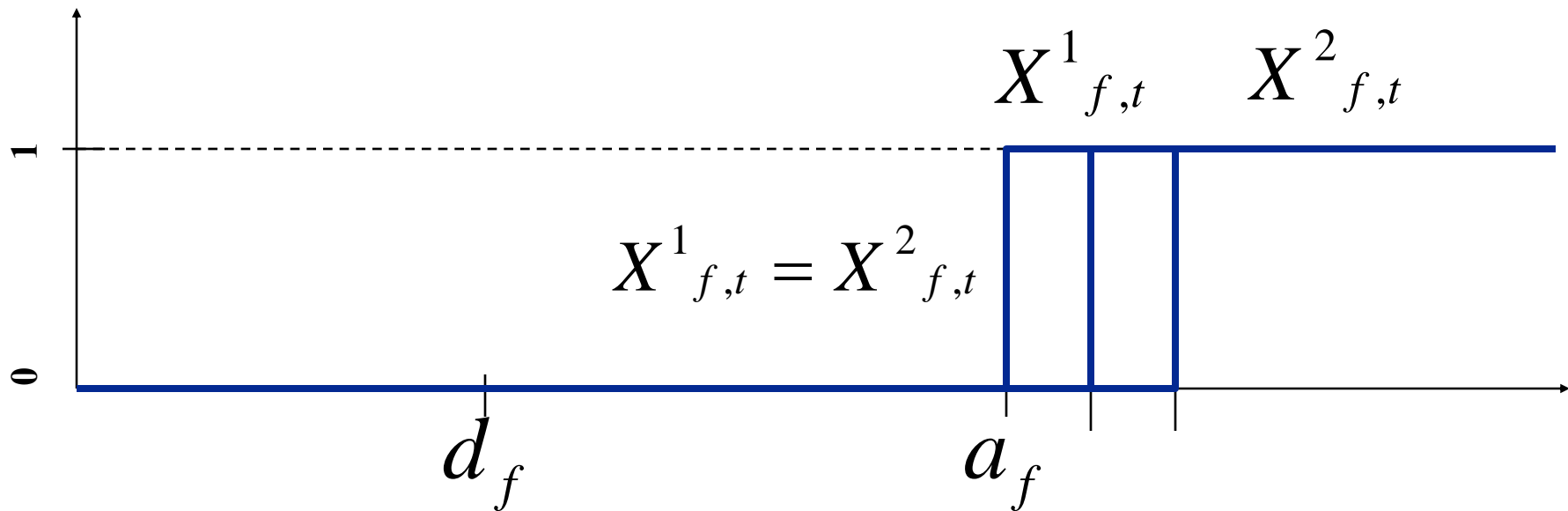
Decision Making Process





Decision Variables

$$X_{f,t}^q = \begin{cases} 1 & \text{if flight } f \text{ is planned to arrive by time period} \\ & t \text{ under scenario } q; \\ 0 & \text{otherwise} \end{cases}$$





Model Formulation

☐ Objective Function

- ☐ Min. Expected Total Cost of Delay (sum of ground and airborne delays)

☐ Major Constraints

- ☐ Number of arrivals during any time interval (period) less than airport capacity
- ☐ Coupling Constraints: decisions cannot be based on a particular scenario until it is completely realized



Model Parameters and Input Data

$\{1..T + 1\}$: set of time periods of uniform duration, T
being the planning horizon

$\Phi = \{1..F\}$: Set of Flights

$Dep_f \in \{1..T\}$: scheduled departure time period of flight f

$Arr_f \in \{1..T\}$: scheduled arrival time period of flight f

λ : Cost ratio between airborne and ground delay



Θ : set of capacity scenarios

P_q : Probability of occurrence of scenario $q \in \Theta$

M_t^q : Airport arrival capacity at time period t
under capacity scenario q

M_{T+1}^q is set to a high value for all $q \in \Theta$



B = total number of branches of the scenario tree; $B \geq |\Theta|$

N_i = number of scenarios represented by i^{th} branch; $i \in \{1..B\}$

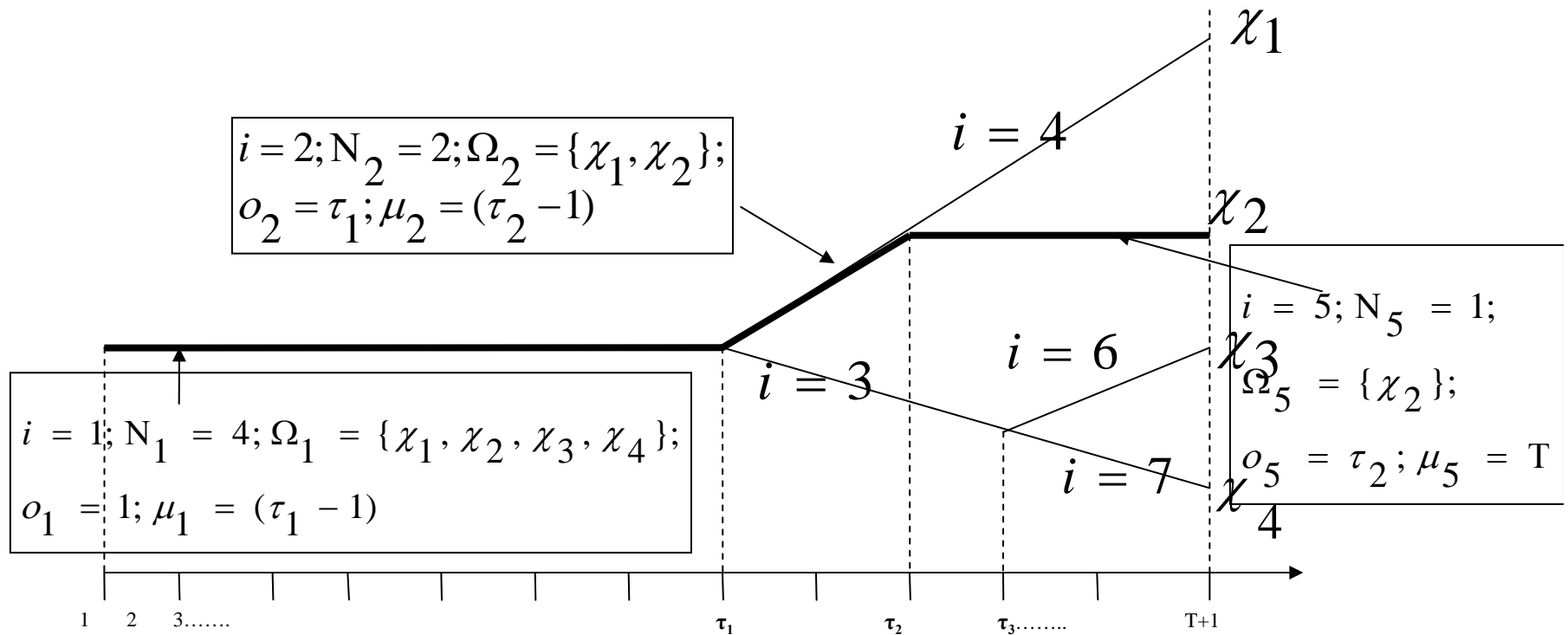
The scenarios represented by branch i is given by set

$$\Omega_i = \{S_1^i, \dots, S_k^i, \dots, S_{N_i}^i\}, S_k^i \in \Theta$$

The time periods corresponding to start and end nodes of a branch are given by o_i and μ_i ; $i \in \{1..B\}$



$$B = 7; \Theta = \{\chi_1, \chi_2, \chi_3, \chi_4\}; |\Theta| = 4$$





Decision Variables

$$X_{f,t}^q = \begin{cases} 1 & \text{if flight } f \text{ is planned to arrive by the end of} \\ & \text{time period } t \text{ under scenario } q; \\ 0 & \text{otherwise} \end{cases}$$

$$\boxed{q \in \Theta, f \in \Phi, \\ t \in \{Arr_f..T + 1\}}$$

$$Y_{f,t}^q = \begin{cases} 1 & \text{if flight } f \text{ is released for departure by the end of} \\ & \text{time period } t \text{ under scenario } q; \\ 0 & \text{otherwise} \end{cases}$$

$$\boxed{q \in \Theta, f \in \Phi, \\ t \in \{Dep_f..T + 1\}}$$

W_t^q = number of aircraft subject to airborne queuing delay
at time t for one or more time periods, under scenario q



Objective Function

$$\text{Min} \sum_{q \in \{1..Q\}} P_q \times \left\{ \left[\sum_{f \in \{1..F\}} \sum_{t=Arr_f}^{T+1} (t - Arr_f) \times (X_{f,t}^q - X_{f,t-1}^q) \right] + \lambda \times \sum_{t=1}^T W_t^q \right\}$$

Constraints

Decision Variables Non Decreasing

$$X_{f,t}^q - X_{f,t-1}^q \geq 0; \quad \forall f \in \Phi, q \in \Theta, t \in \{Arr_f .. T + 1\}$$

Planned Departure Time of Flights

$$Y_{f,t}^q = \begin{cases} X_{f,t+Arr_f}^q - Dep_f & \text{if } t + Arr_f - Dep_f \leq T \\ 1 & \text{otherwise} \end{cases}$$



Arrival Capacity

$$W_{t-1}^q - W_t^q + \sum_{f \in \Phi} \left(X_{f,t}^q - X_{f,t-1}^q \right) \leq M_t^q; \quad t \in \{1..T+1\}, q \in \Theta$$

Feasibility Conditions

$$W_0^q = W_{T+1}^q = 0$$

$$X_{f,T+1}^q = 1 \quad \forall f \in \Phi, q \in \Theta$$

Coupling Constraints for Ground Holding Decision Variables

$$Y_{f,t}^{S_1^i} = \dots = Y_{f,t}^{S_k^i} = \dots = Y_{f,t}^{S_{N_i}^i}; \quad f \in \Phi, t \in \{1..T\}; S_k^i \in \Omega_i : N_i \geq 2 \text{ and } o_i \leq t \leq \mu_i$$



Static vs Dynamic Formulation

- ❑ Static Stochastic Model (Ball et al 2003, Richetta-Odoni 1993) is a special case of dynamic model.
- ❑ The decisions are taken once at the beginning of day and not revised later.

$X_{f,t}^q$ is same for all $q \in \Theta$; i.e., superscript q in the decision variables can be dropped.

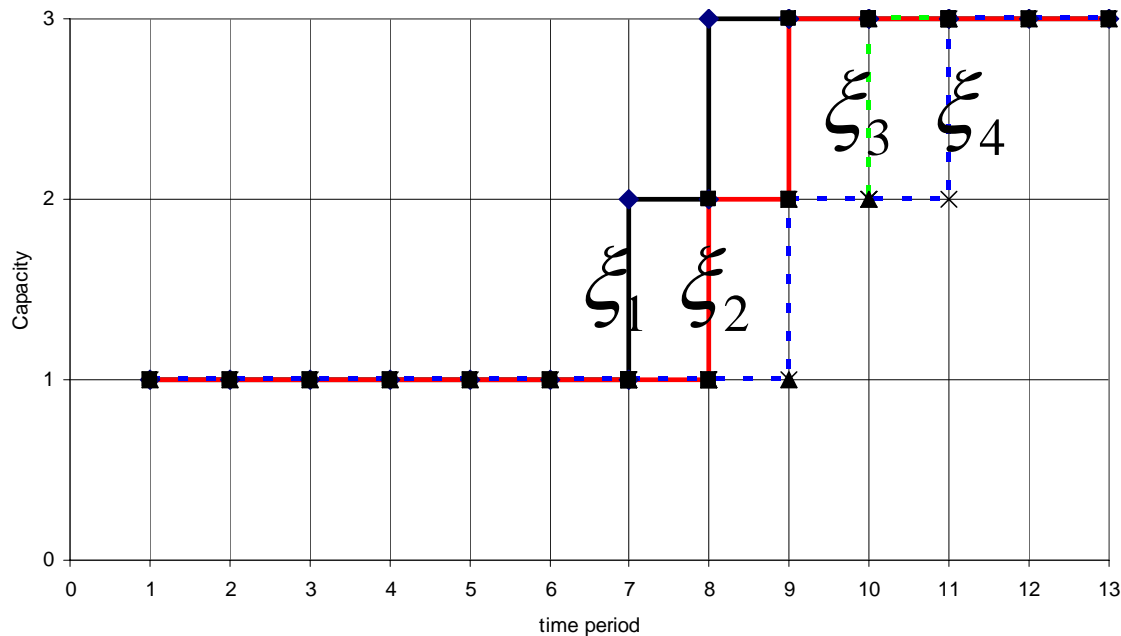
Therefore, $X_{f,t}^q$ can be denoted as $X_{f,t}; \forall q \in \Theta$

Similarly, $Y_{f,t}^q = Y_{f,t}; \forall q \in \Theta$



Example

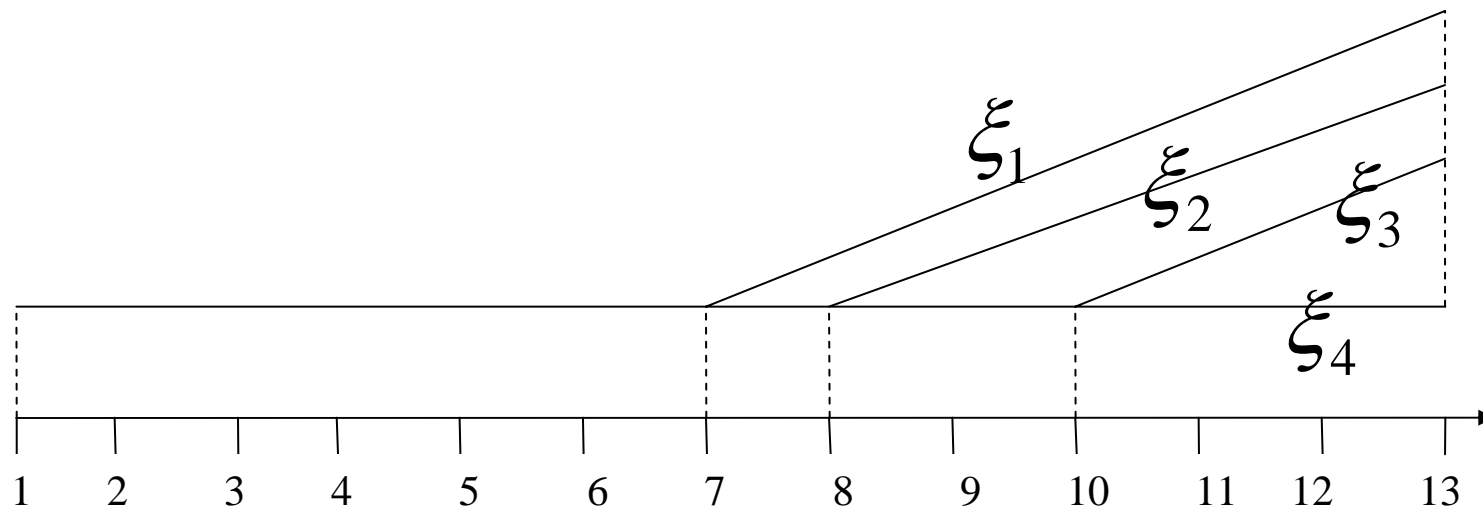
- 4 Scenarios, 4 Decision Stages
- 13 Time Periods
- 13 Flights
- Cost Ratio $\lambda = 5$



Probability Mass Function : $P\{\xi_1\} = 0.5; P\{\xi_2\} = 0.3; P\{\xi_3\} = 0.1; P\{\xi_4\} = 0.1$



Scenario Tree

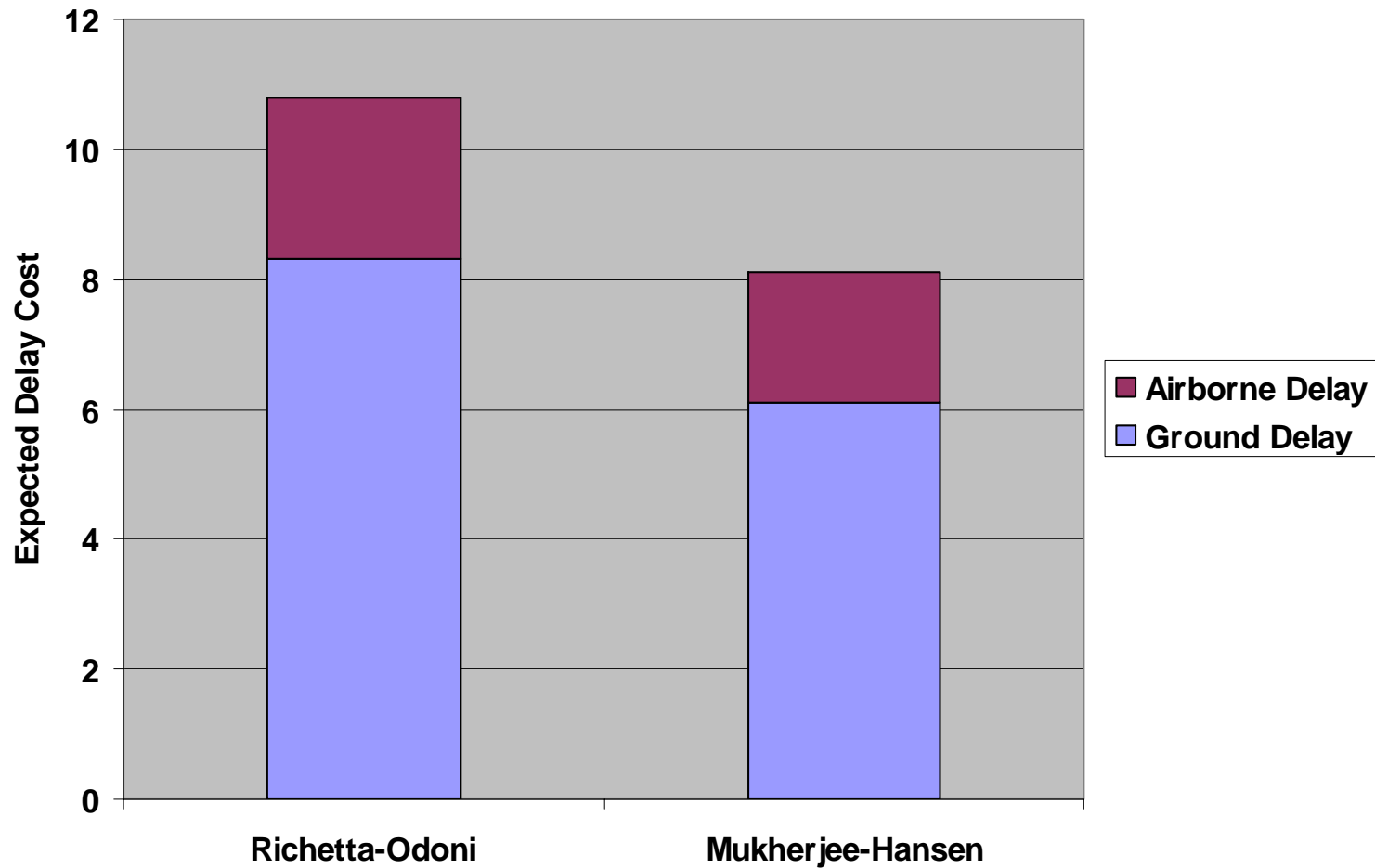




Flight No.	Dep.	Arr.	Decision Stage	Richetta-Odoni Model				Mukherjee-Hansen Model (Optimal Solution 1)				Mukherjee-Hansen Model (Optimal Solution 2)			
				ξ_1	ξ_2	ξ_3	ξ_4	ξ_1	ξ_2	ξ_3	ξ_4	ξ_1	ξ_2	ξ_3	ξ_4
2	6	7	1	3	3	3	3	1	2	5	5	1	2	4	4
3	2	8	1	0	0	0	0	1	1	1	1	1	1	1	1
7	5	9	1	0	0	0	0	0	0	0	0	0	0	0	0
8	7	9	2	0	3	3	3	0	1	2	2	0	1	3	3
12	9	11	3	0	0	1	1	0	0	1	1	0	0	1	1



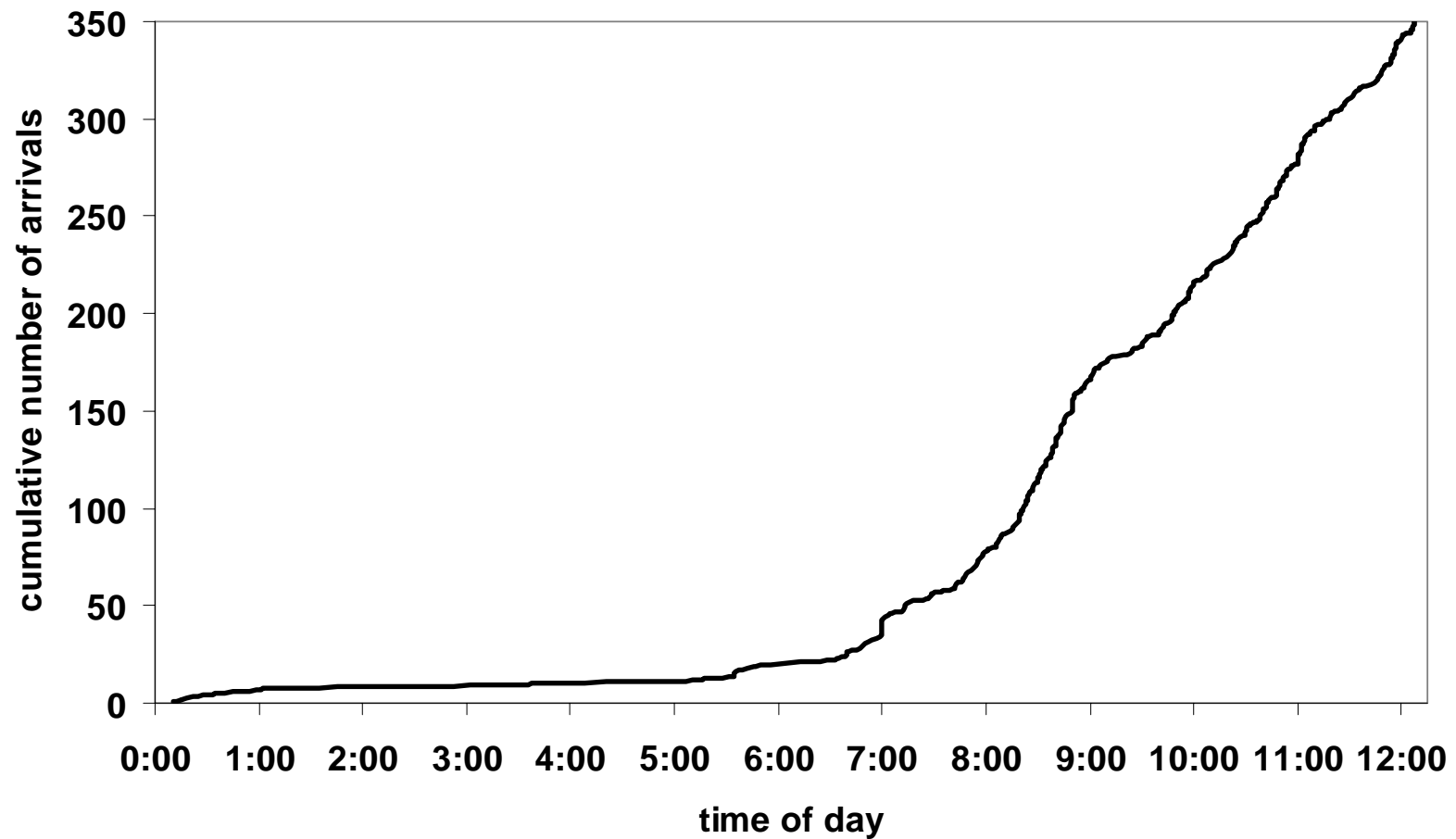
Expected Cost of Delay





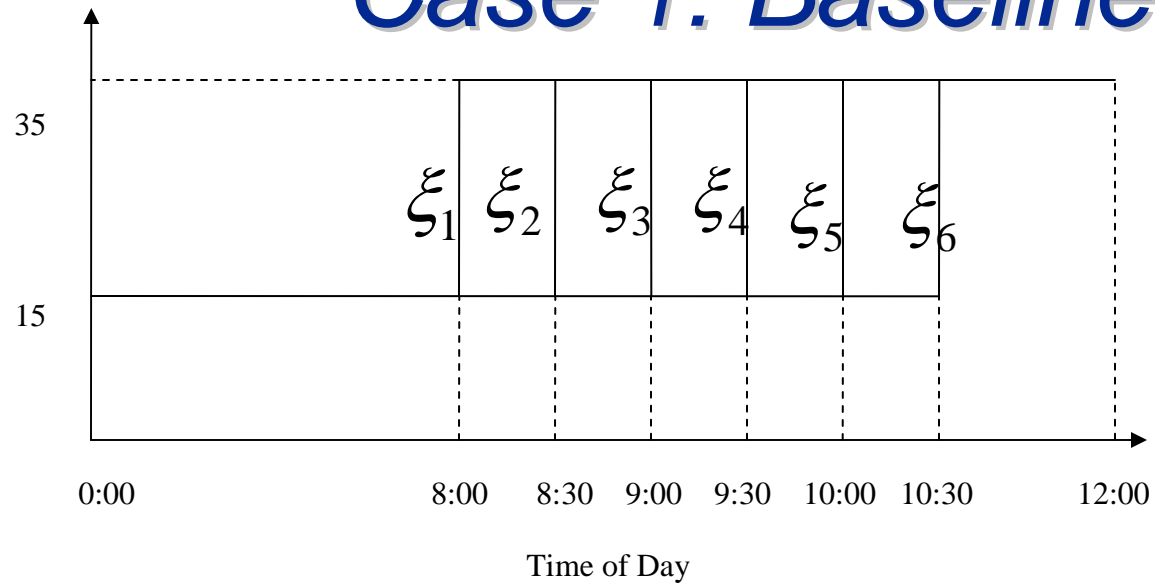
Cumulative Arrivals at DFW

July14_2003





Case 1: Baseline



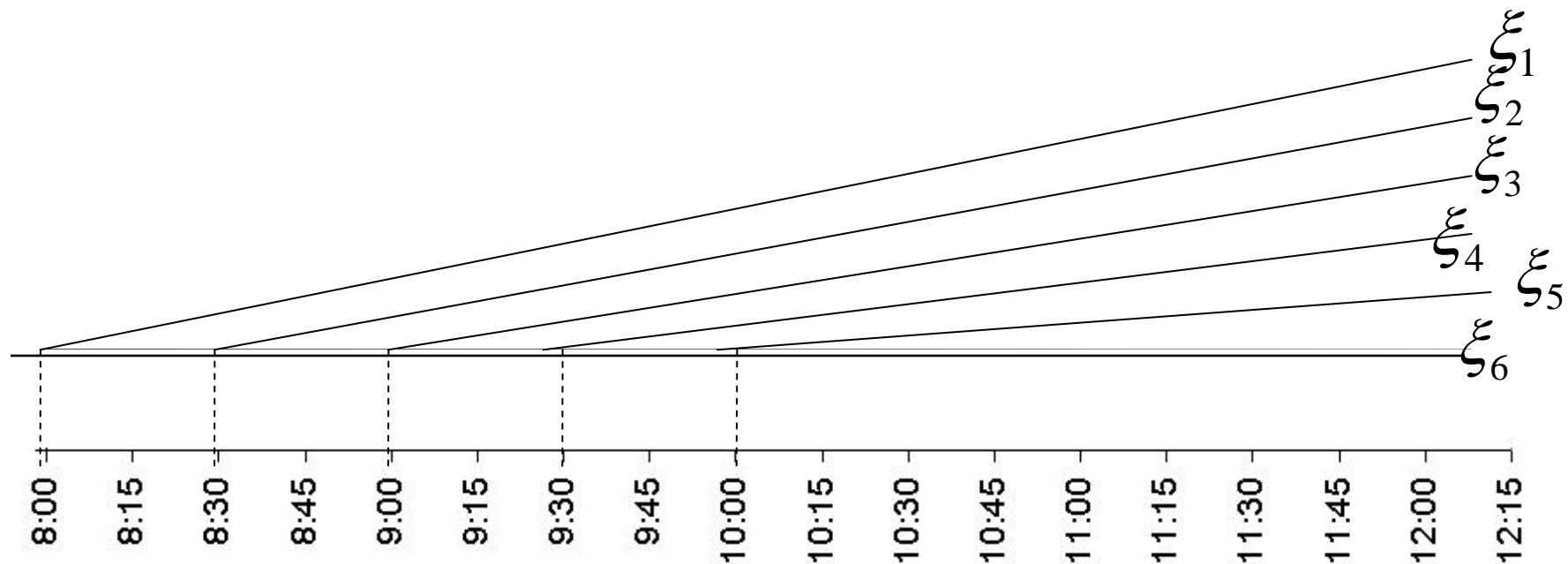
Probability Mass Function

$$P\{\xi_1\} = 0.4; P\{\xi_2\} = 0.2; P\{\xi_3\} = 0.1; P\{\xi_4\} = 0.1; P\{\xi_5\} = 0.1; P\{\xi_6\} = 0.1$$

Cost Ratio $\lambda = 3$



Scenario Tree for *Baseline Case*





Other Cases

Case 2: Change in Cost Ratio. $\lambda = 25$

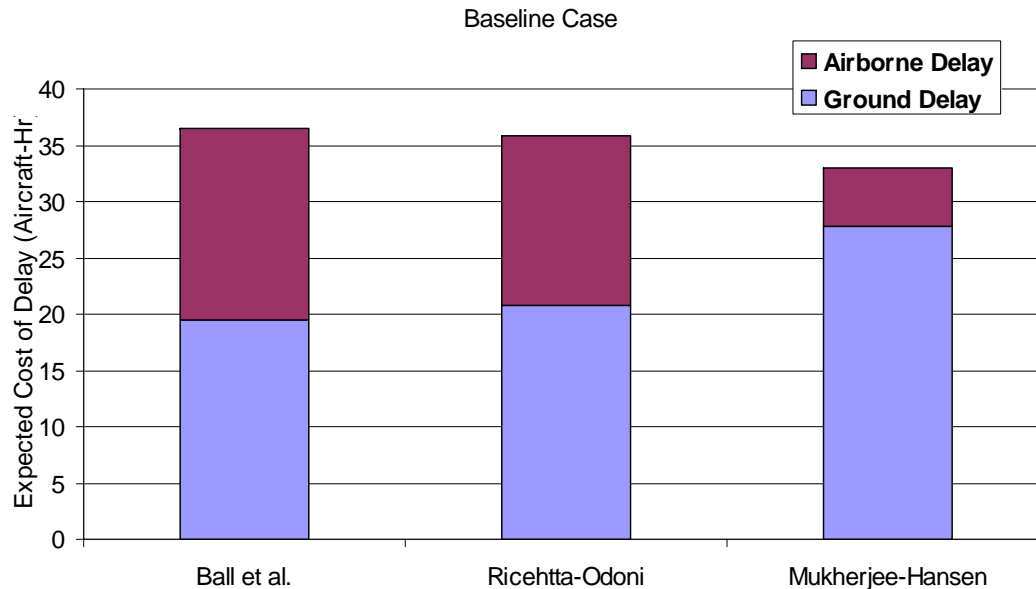
Case 3: Change in PMF

$$P\{\xi_1\} = 0.1; P\{\xi_2\} = 0.1; P\{\xi_3\} = 0.1; P\{\xi_4\} = 0.1; P\{\xi_5\} = 0.2; P\{\xi_6\} = 0.4$$

Case 4: Early Branching. Scenarios are realized 30 minutes earlier



Results: Baseline Case



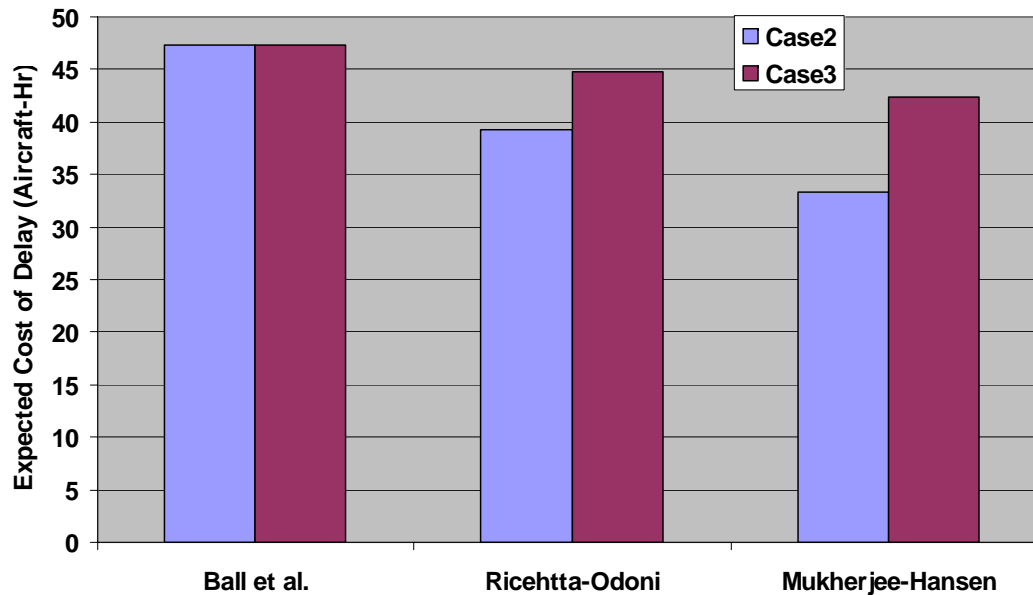
- ❑ Mukherjee-Hansen Model
 - ❑ Ground delays more severe
 - ❑ Less airborne delays
 - ❑ Total expected cost least
- ❑ Delay reduction compared to Static Model
 - ❑ 10% in Mukherjee-Hansen Model
 - ❑ 2% in Richetta-Odoni



Planned Arrival Rates in Baseline Case												
Time Period	Mukherjee-Hansen Model						Richetta-Odoni Model					
	ξ_1	ξ_2	ξ_3	ξ_4	ξ_5	ξ_6	ξ_1	ξ_2	ξ_3	ξ_4	ξ_5	ξ_6
9:00AM-9:15AM	33	25	25	25	25	25	35	34	34	34	34	34
9:15 AM-9:30 AM	16	5	5	5	5	5	14	14	14	14	14	14
9:30AM-9:45AM	12	26	18	18	18	18	12	12	12	12	12	12
9:45AM-10:00AM	20	25	12	12	12	12	20	21	13	13	13	13
10:00AM-10:15AM	12	12	33	33	33	33	12	12	20	20	20	20



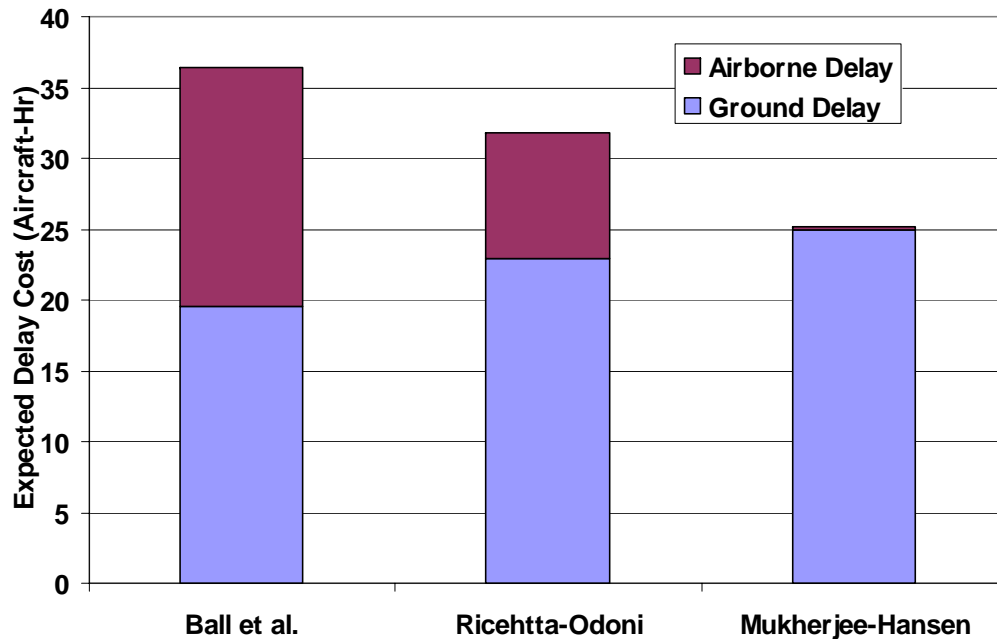
Cases 2 and 3



- ☐ No airborne delays
- ☐ Static model plans for worst scenario
- ☐ Dynamic models adaptive to changing conditions
- ☐ Delays are higher in case 3 due to high probability of worse conditions



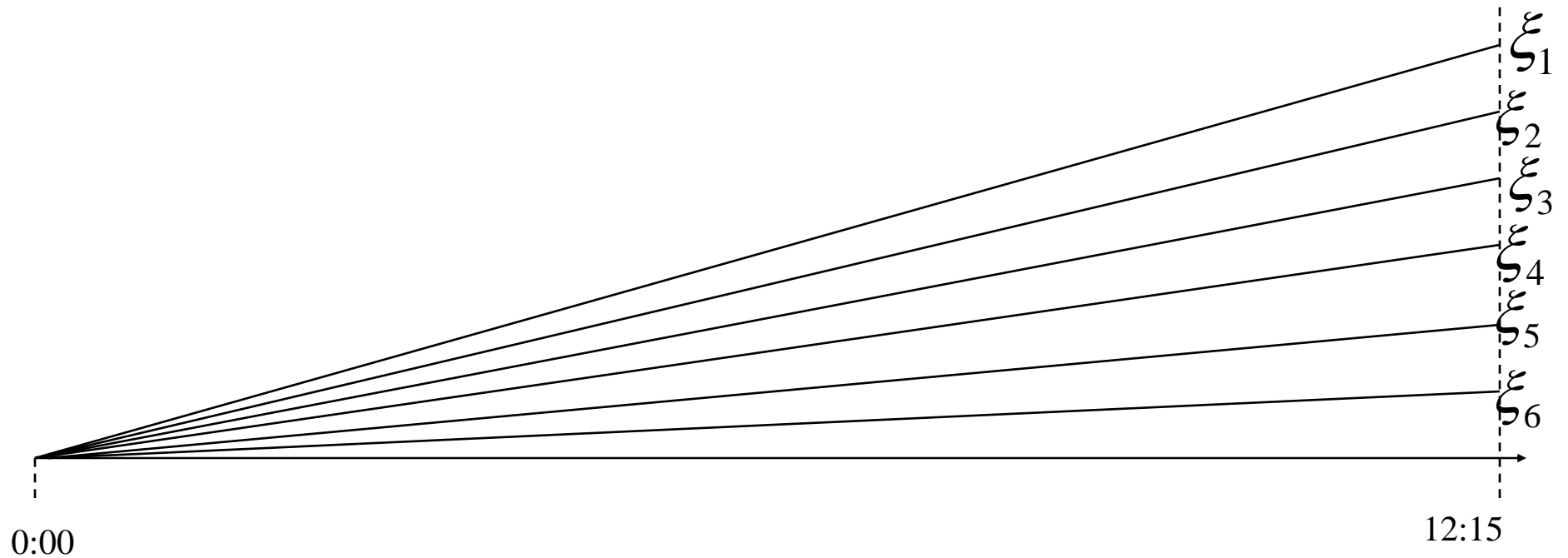
Case 4: 30 Minutes Early Information

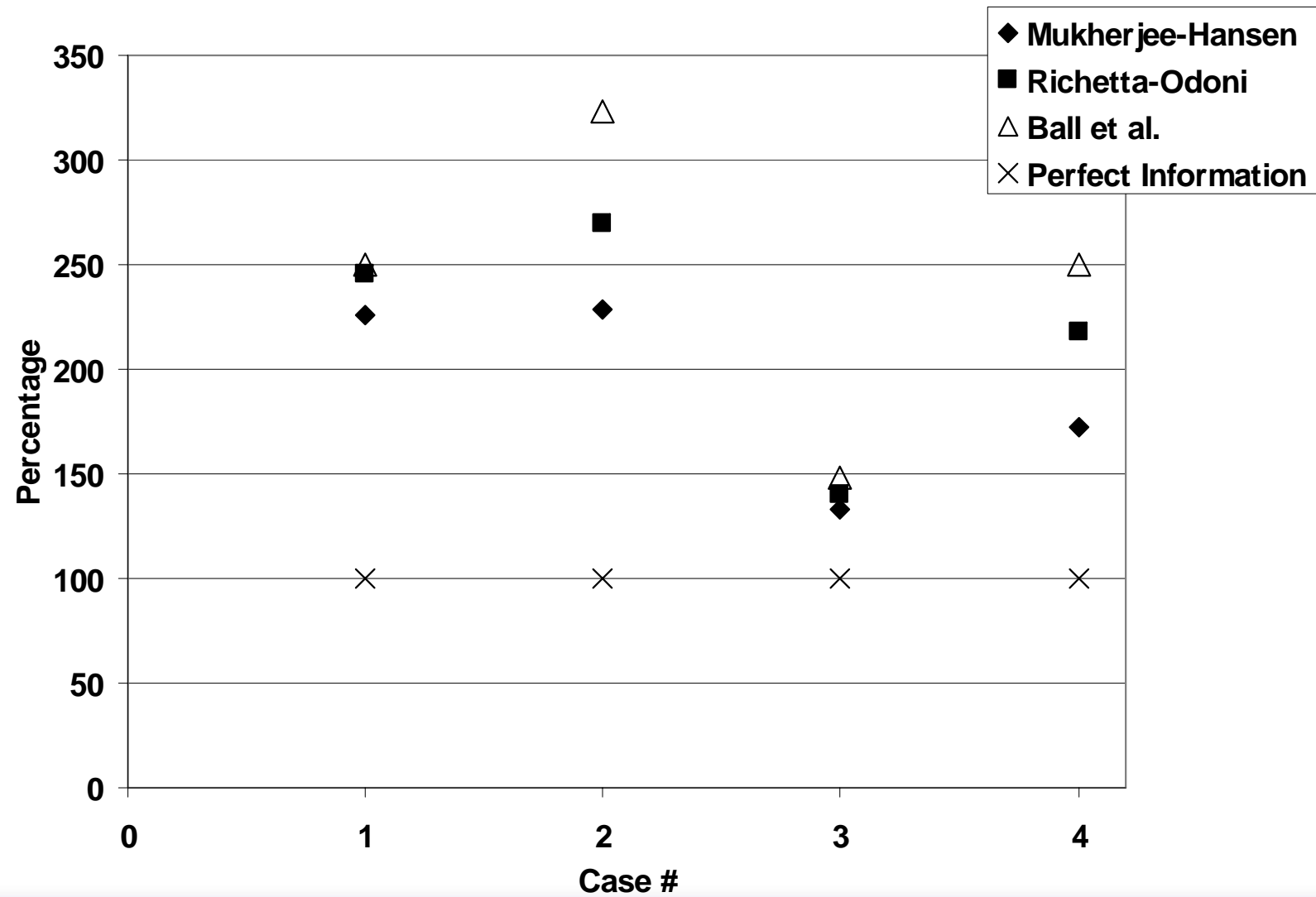


- ❑ Static model produces same delays as in baseline case
- ❑ Value of early information
- ❑ Mukherjee-Hansen Model
 - ❑ Least delays
 - ❑ Absorbs most of the delay by ground holding



Perfect Information Case







Alternative Objective Functions

Minimizing Expected Squared Deviation from RBS Allocation

$$\text{Min}_{q \in \{1..Q\}} \sum P\{q\} \times \left\{ \left[\sum_{f \in \{1..F\}} \sum_{t=Arr_f}^{T+1} \left(t - RBS_f^q \right)^2 \times (X_{f,t}^q - X_{f,t-1}^q) \right] + \lambda \times \sum_{t=1}^T W_t^q \right\}$$



Multi-Criteria Optimization

$$\text{Min}_{q \in \{1..Q\}} \sum P\{q\} \times \left\{ \begin{aligned} & \left[\sum_{f \in \Phi} \sum_{t=Arr_f}^{T+1} (t - Arr_f) \times (X_{f,t}^q - X_{f,t-1}^q) \right] + \\ & \text{weight} \times \left[\sum_{f \in \Phi} \sum_{t=Arr_f}^{T+1} \left(t - RBS_f^q \right)^2 \times (X_{f,t}^q - X_{f,t-1}^q) \right] + \\ & \lambda \times \sum_{t=1}^T W_t^q \end{aligned} \right\}$$



Work in Progress

- ☐ Reformulating the model as a minimum cost network flow problem.
- ☐ Ability to handle time varying unconditional probabilities of the capacity scenarios



Acknowledgements

- ☐ Authors are thankful to Prof. Mike Ball for his thoughtful suggestions on this research.