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# Modeling Demand Uncertainties During Ground Delay Programs

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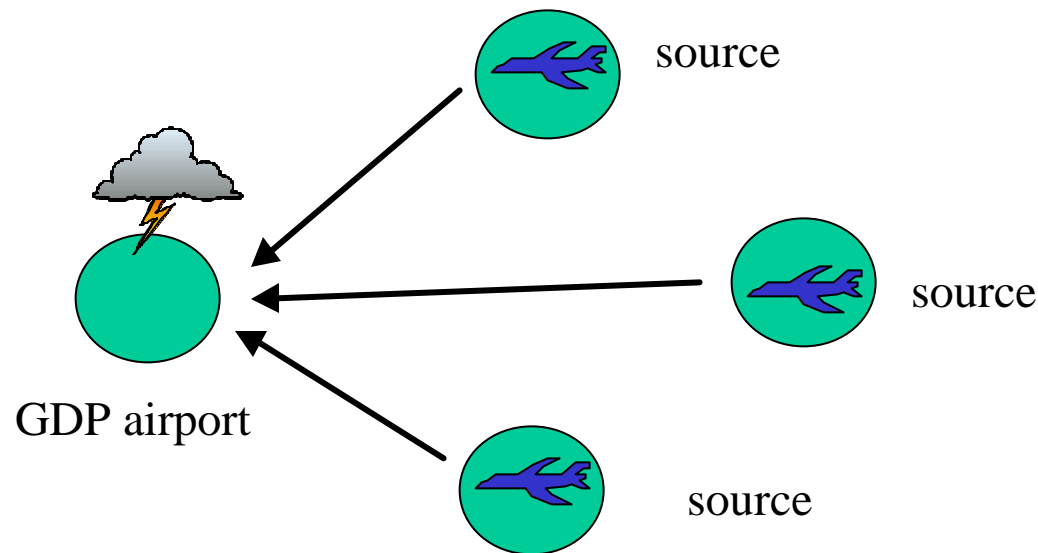
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# Ground Delay Programs (GDPs)

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- To balance arrival demand and capacity at the afflicted airport
- Transfer costly airborne delay to less expensive ground delay



- Input parameters - Airport Capacity and Arrival Demand
- Capacity and demand , both stochastic in nature

# Demand Uncertainties

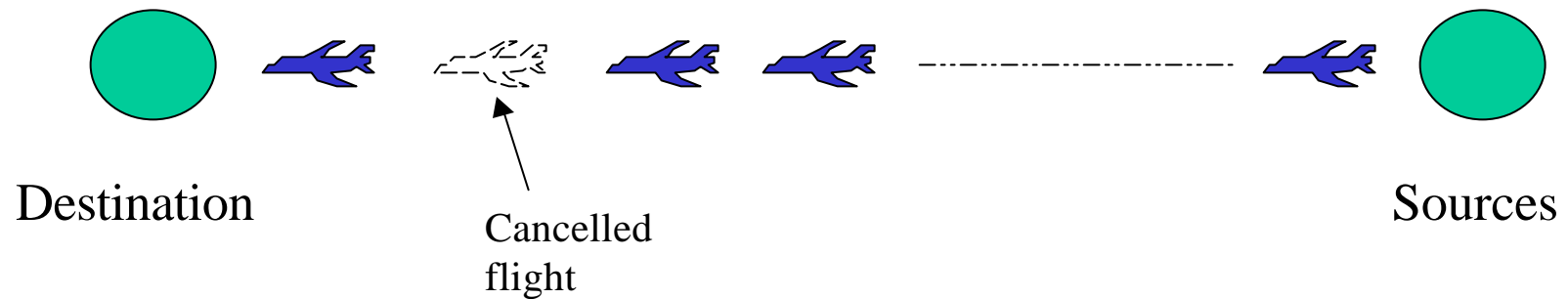
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- ➔ Three main sources of demand uncertainties
  - Flight Drifts,
  - Flight Cancellations, and
  - Pop-up Flights
  
- ➔ Combined Effects
  - Under-Utilization of the airport resources (slots)
  - Unpredictable arrival sequence
  - Increased airborne holding

# Flight Cancellations

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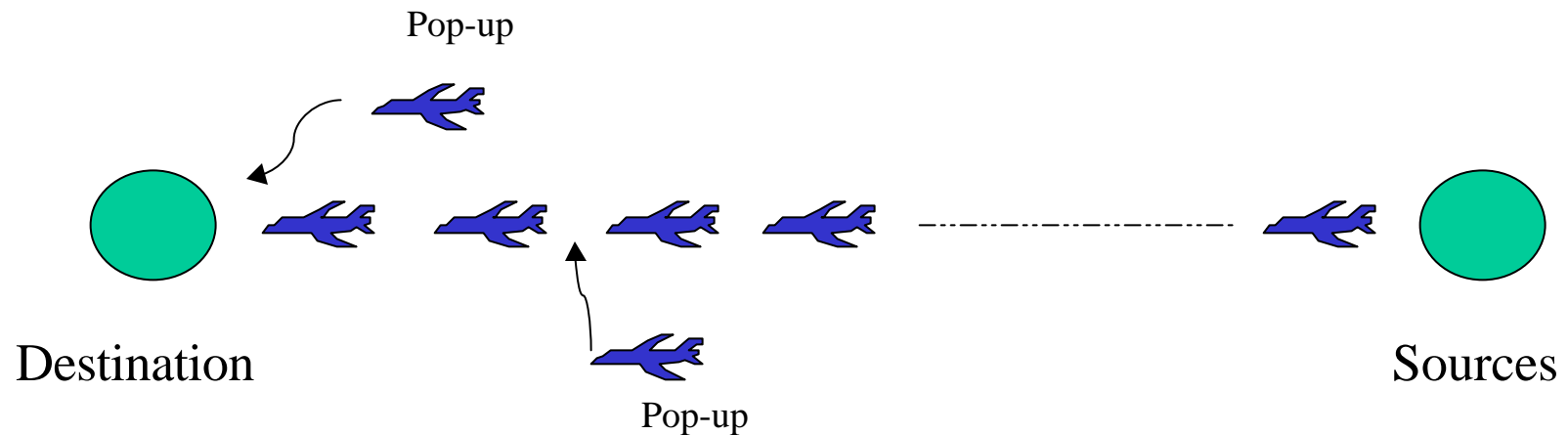
- Flight Cancellations without notices in advance, cause “holes” in the arrival sequence
- Timed Out (TO) Cancellations almost always result in slots being unused



# Pop-up Flights

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- ➔ Any flight that arrived during a GDP and that first appeared in the ADL after the GDP model time

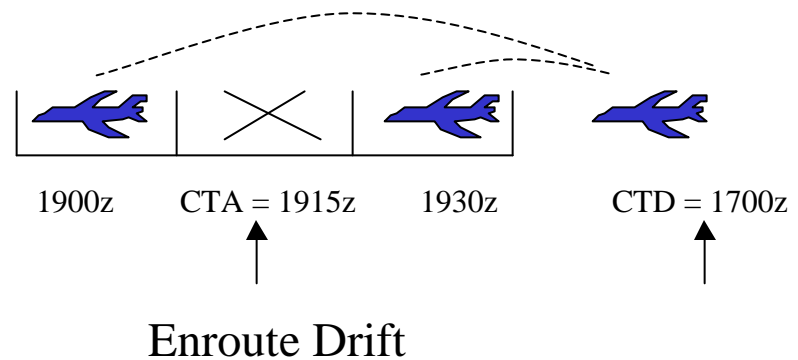
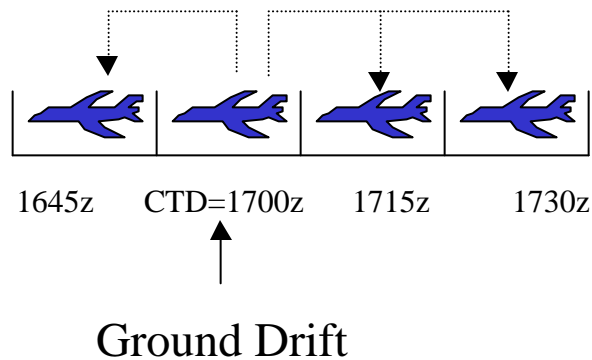


- ➔ Pop-ups add to the arrival demand and displace the actual arrival sequence

# Flight Drifts

→ Flight Drifts are results of :

- CTD non-compliance , where ARTD > or < CTD
- CTA non-compliance , where AETE > or < OETE



→ Net Drift = Ground Drift + Enroute Drift

- ARTA - Actual Runway Time of Arrival
- ARTD - Actual Runway Time of Departure
- AETE - Actual Enroute Time

- CTA - Control Time of Arrival
- CTD - Control Time of Departure
- OETE - Original Estimated Enroute Time

# Modeling Demand Uncertainties

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- ➔ Stochastic Mixed Integer Optimization (SMIO) Model
  - Incorporates only flight cancellations and pop-ups
  - Generates Optimal Planned Arrival Rates (PAARs) for any GDP scenario
  
- ➔ Simulation Model
  - Incorporates flight cancellations, pop-ups and drifts
  - Validates the SMIO model by generating Pareto Optimal PAARs for a large set of scenarios

# Details on SMIO Model

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- ➔ Objective Function : Minimize the expected airborne queue
- ➔ Variables :  $X_{\text{paar}}(k,t) = 1$  if PAAR =  $k$  in time period  $t$ ; else 0  
 $Y(k,j,t)$  : probability that at the end of time period “ $t$ ” an airborne queue of size “ $j$ ” exists and PAAR =  $k$  in time period “ $t$ ”
- ➔ Main Input Parameters : AAR( $t$ ),  $P_{\text{cnx}}$ ,  $P_{\text{pop}}$ , and Utilization Parameter “ $\epsilon$ ”
- ➔ Main Constraints
  - **Markovian Constraint** :  $\sum_k Y(k,j,t) = \sum_i q(k,i,j,t) Y(k,i,t-1)$ ,  
where  $q(k,i,j,t) = \Pr\{i + \text{number of arrivals in } t - \text{AAR}(t) = j \mid \text{PAAR} = k\}$
  - **Utilization Constraint** : Expected Number of unutilized slots  $\leq \epsilon$
- ➔ Outputs : Optimal PAARs and Optimal Expected Queue Size



# Formulation of SMIO Model

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Minimize  $\sum_t \sum_k \sum_j j Y(k,j,t)$

Subject to:

$$\sum_k X_{\text{paar}}(k,t) = 1 \quad \forall t = 1, 2, \dots, P \quad \dots\dots\dots (1)$$

$$\sum_j Y(k,j,t) \leq X_{\text{paar}}(k,t) \quad \forall j = 1, 2, \dots, \text{MaxQ} \quad \dots\dots\dots (2)$$

$$\sum_{k^*} Y(k^*,j,t) \leq \sum_k \sum_i q(k,i,j,t) Y(k,i,t-1) \quad \forall j, \forall t \quad \dots\dots\dots (3)$$

$$\sum_t \sum_k \sum_i Q_e(k,i,t) Y(k,i,t-1) \leq \varepsilon \quad \dots\dots\dots (4)$$

$$X_{\text{paar}}(k,t) \in \{0,1\}$$

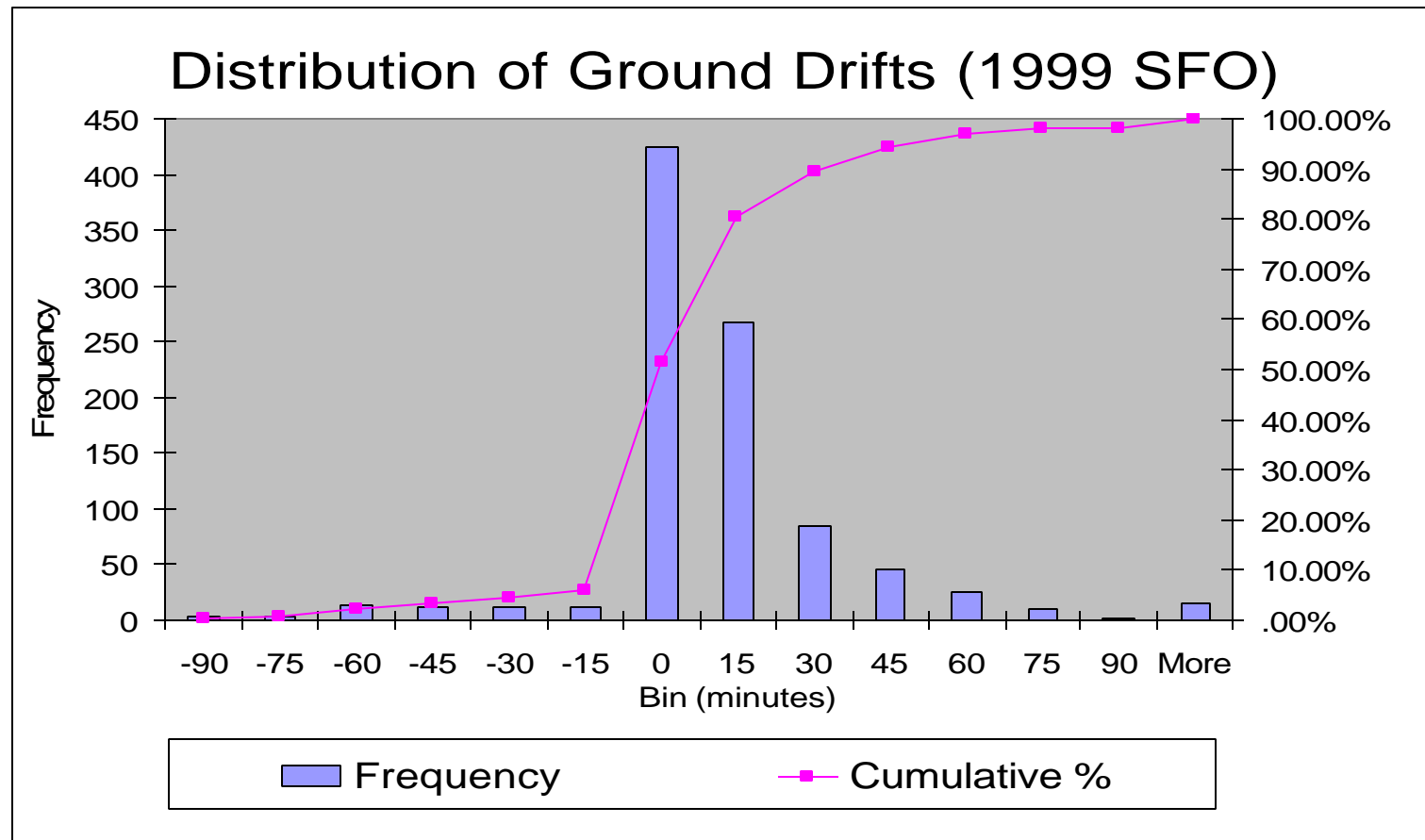
# Details on Simulation Model

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- ➔ Single-Server Queuing Model
  
- ➔ Input Distributions
  - Geometric distribution for flight cancellations
  - Empirical distribution for drifts
  - Exponential distribution for pop-ups
  
- ➔ Performance Measures
  - Ground delay
  - Airborne delay
  - Utilization

# Empirical Analysis of Drifts

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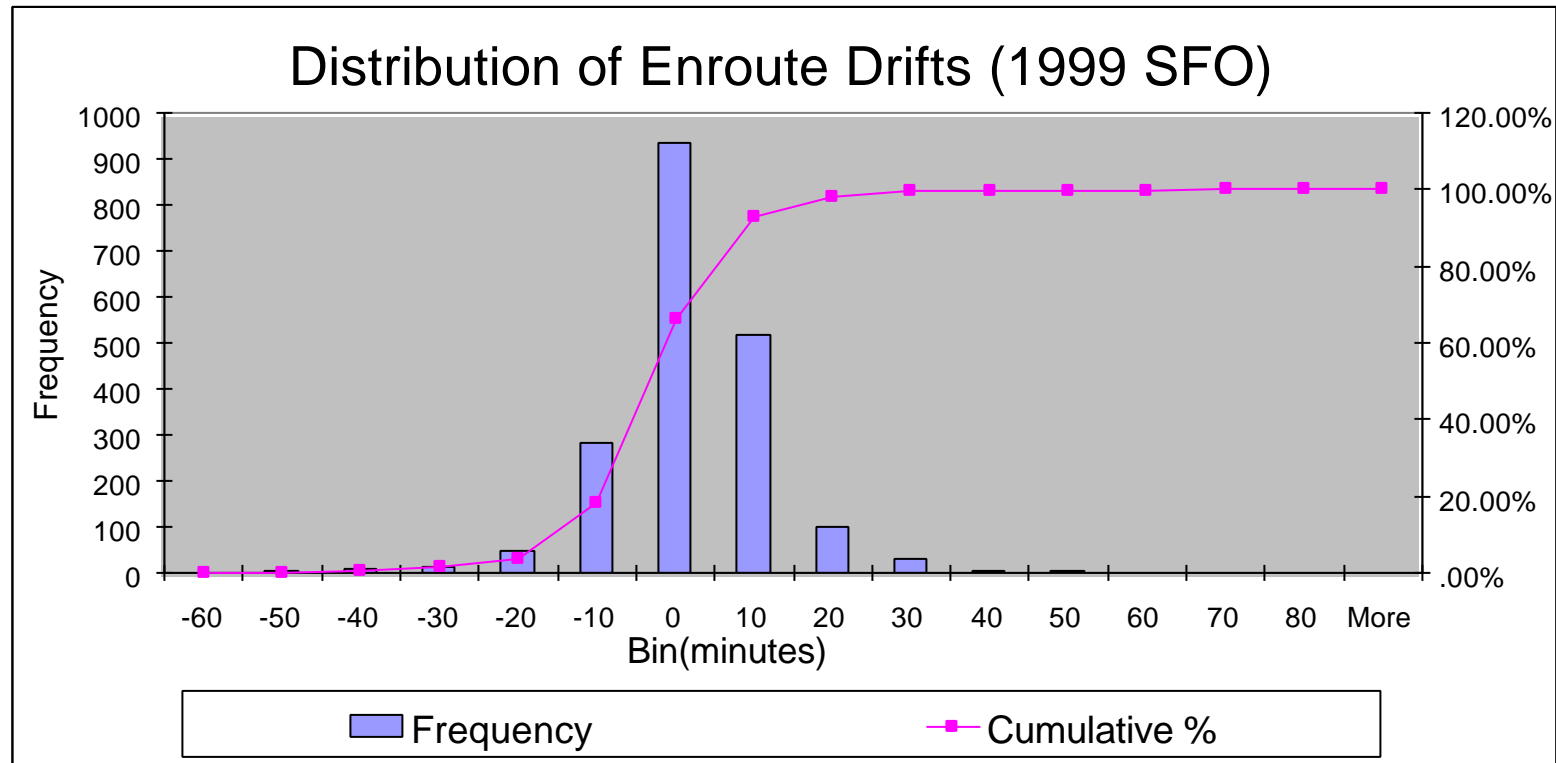


✈ Ground Drift = ARTD - CTD

✈ Mean is shifted to the right - more forward drifts

# Empirical Analysis of Drifts (contd..)

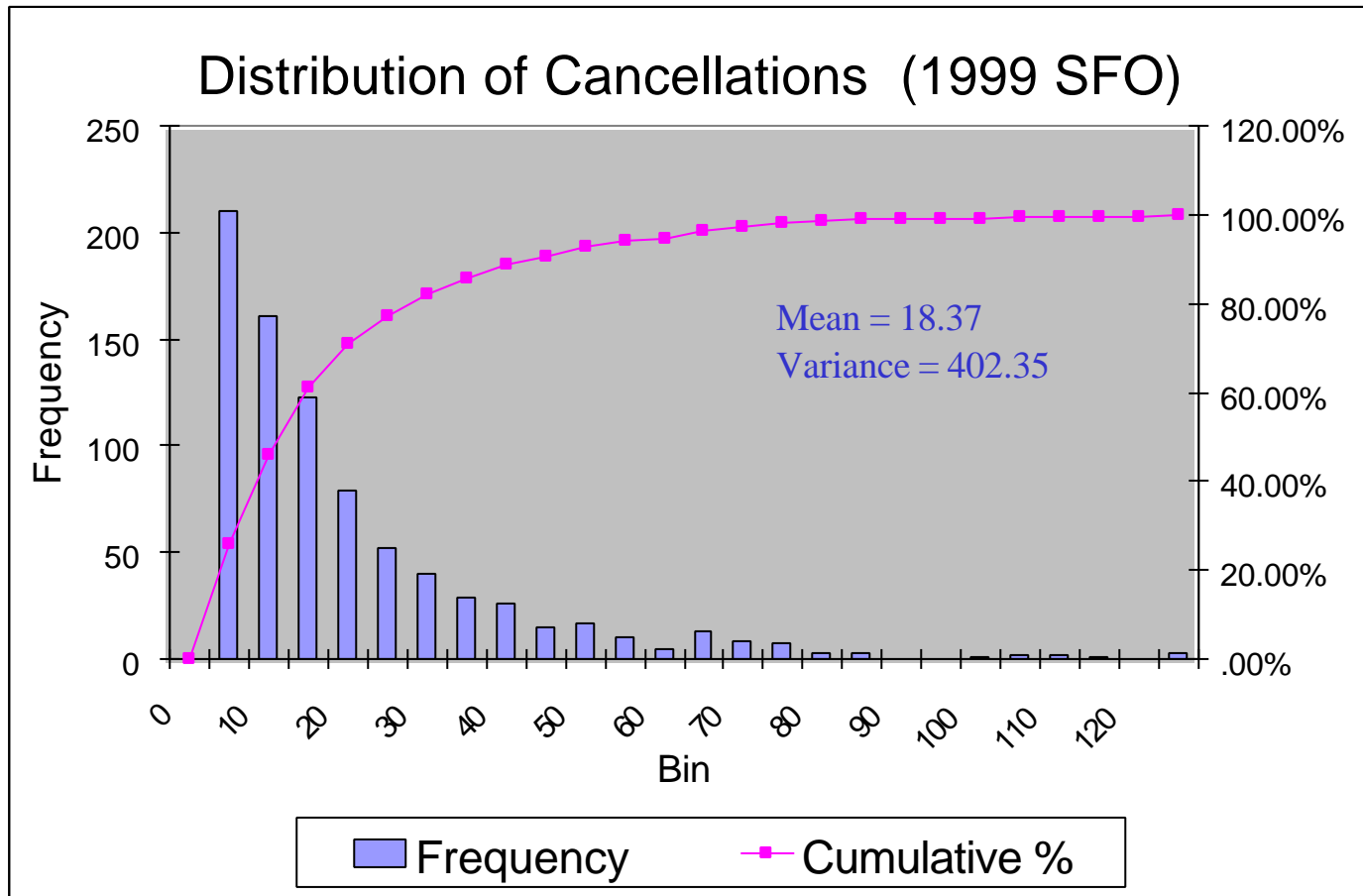
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- ➔ Enroute Drift = AETE - OETE
- ➔ Actual Enroute Time less than expected
- ➔ Enroute Drifts confined to a small window

# Empirical Analysis of Cancellations

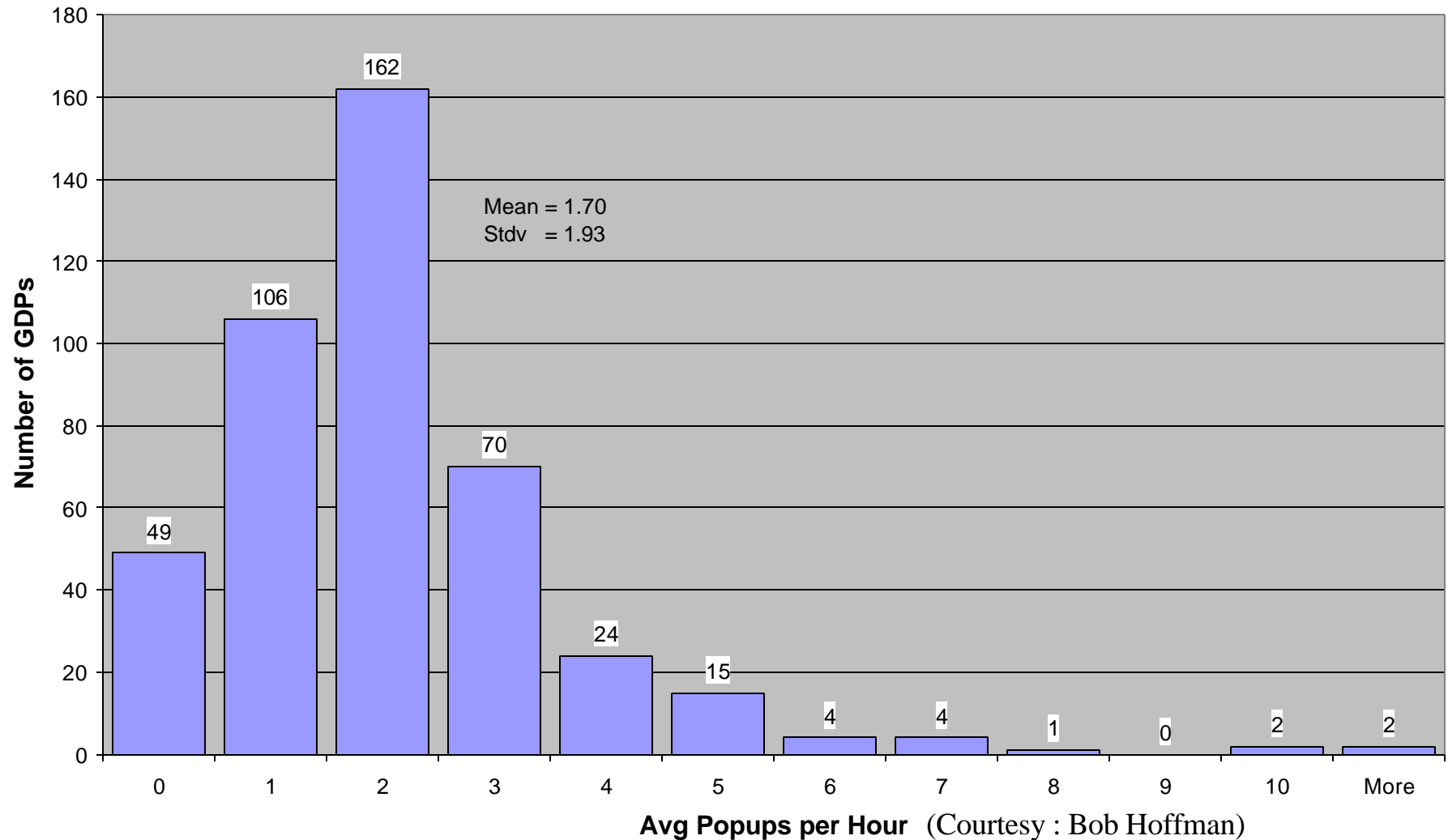
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- Cancellations follow a geometric distribution during GDP

# Empirical Analysis of Pop-up Flights

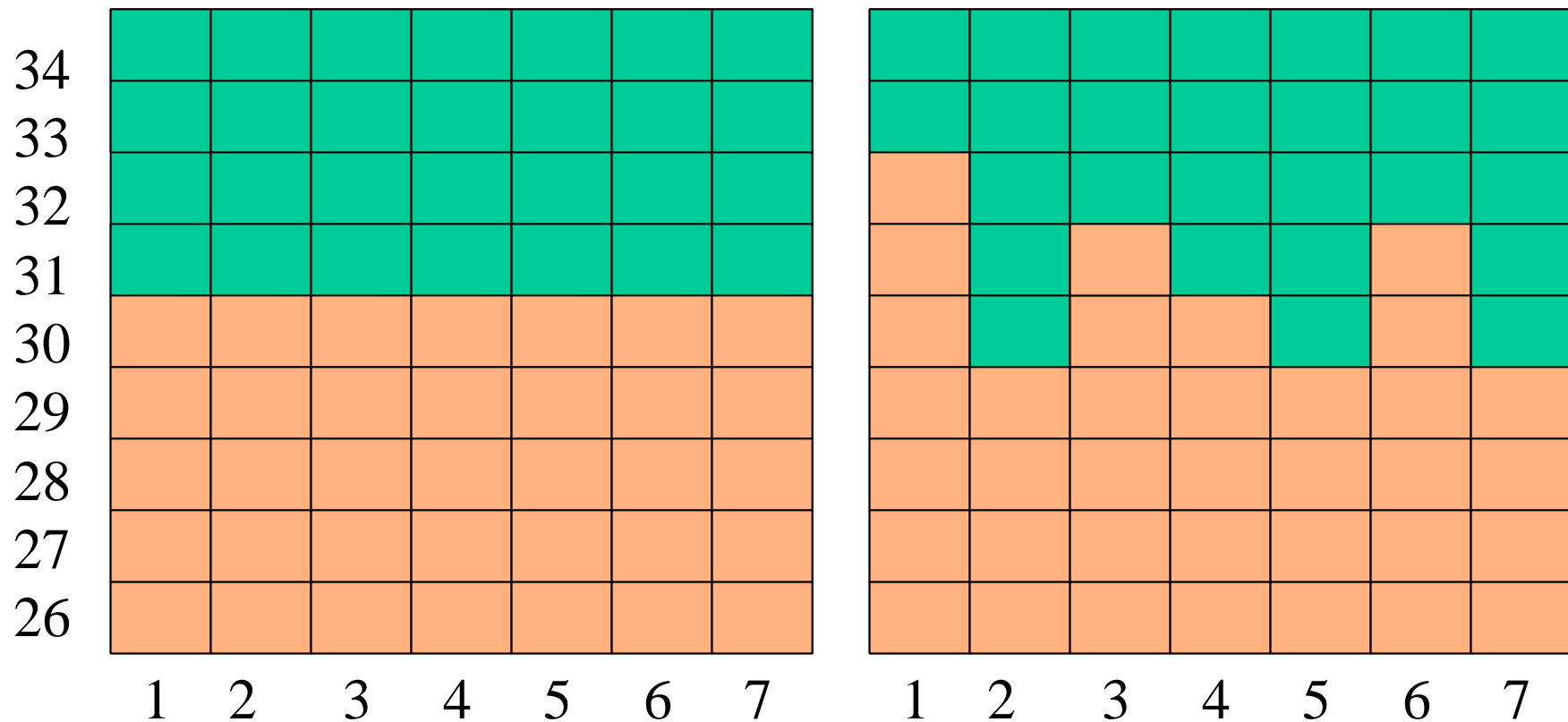
GDP Avg Popup per Hour  
SFO



# Results for SMIO Model

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→ Capacity Scenario : (30,30,30,30,30,30,30) on 05/01/98 SFO



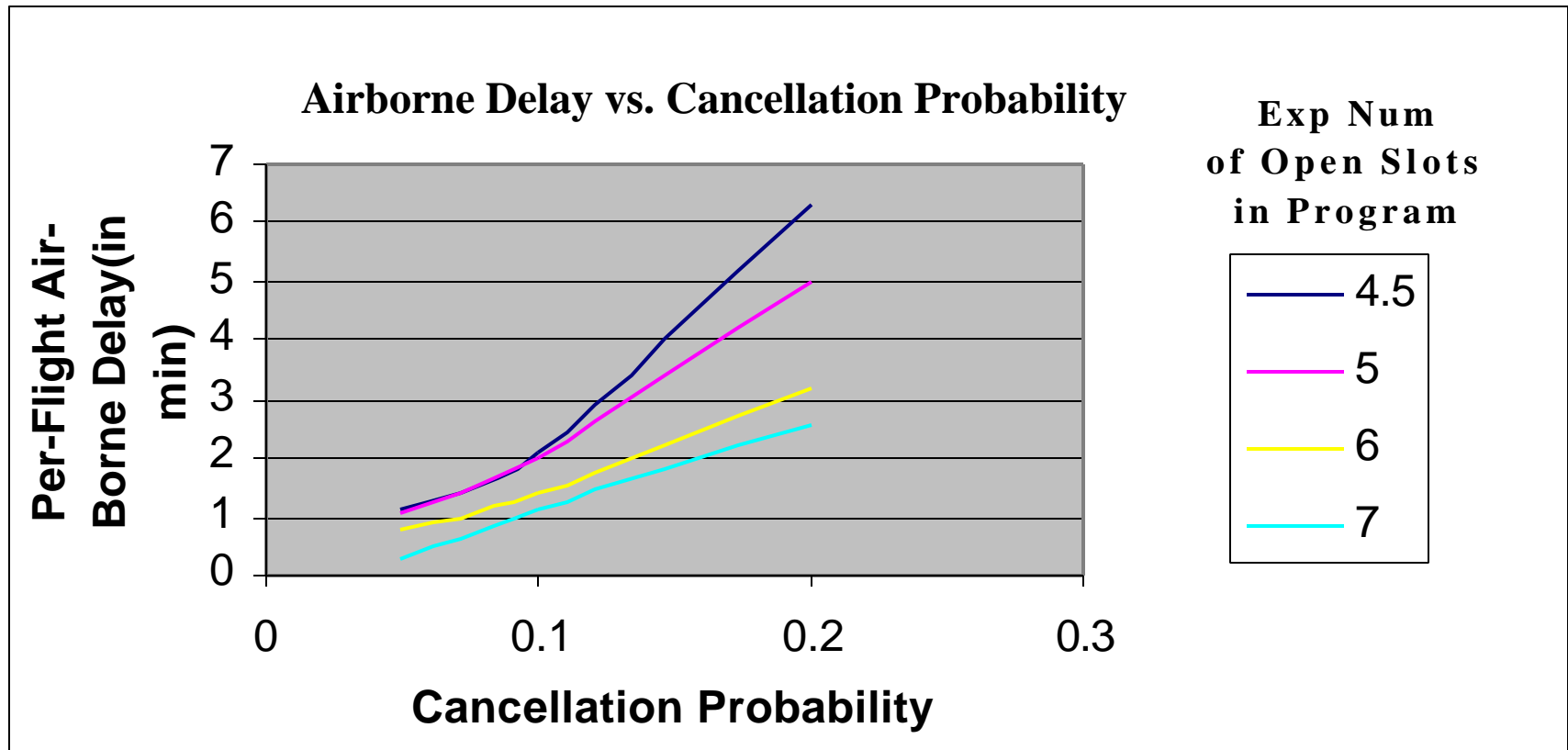
Time Period  
(hour)

**ATC PAARs**

**SMIO PAARs**

# Results for SMIO Model (contd.)

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# Results for Simulation Model

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- Capacity Scenario : (30,30,30,30,30,30)
- Tested the scenarios for all PAARs in the interval [28 34]
- Used **Pareto Optimality** with Airborne Delay and Utilization as Objective functions

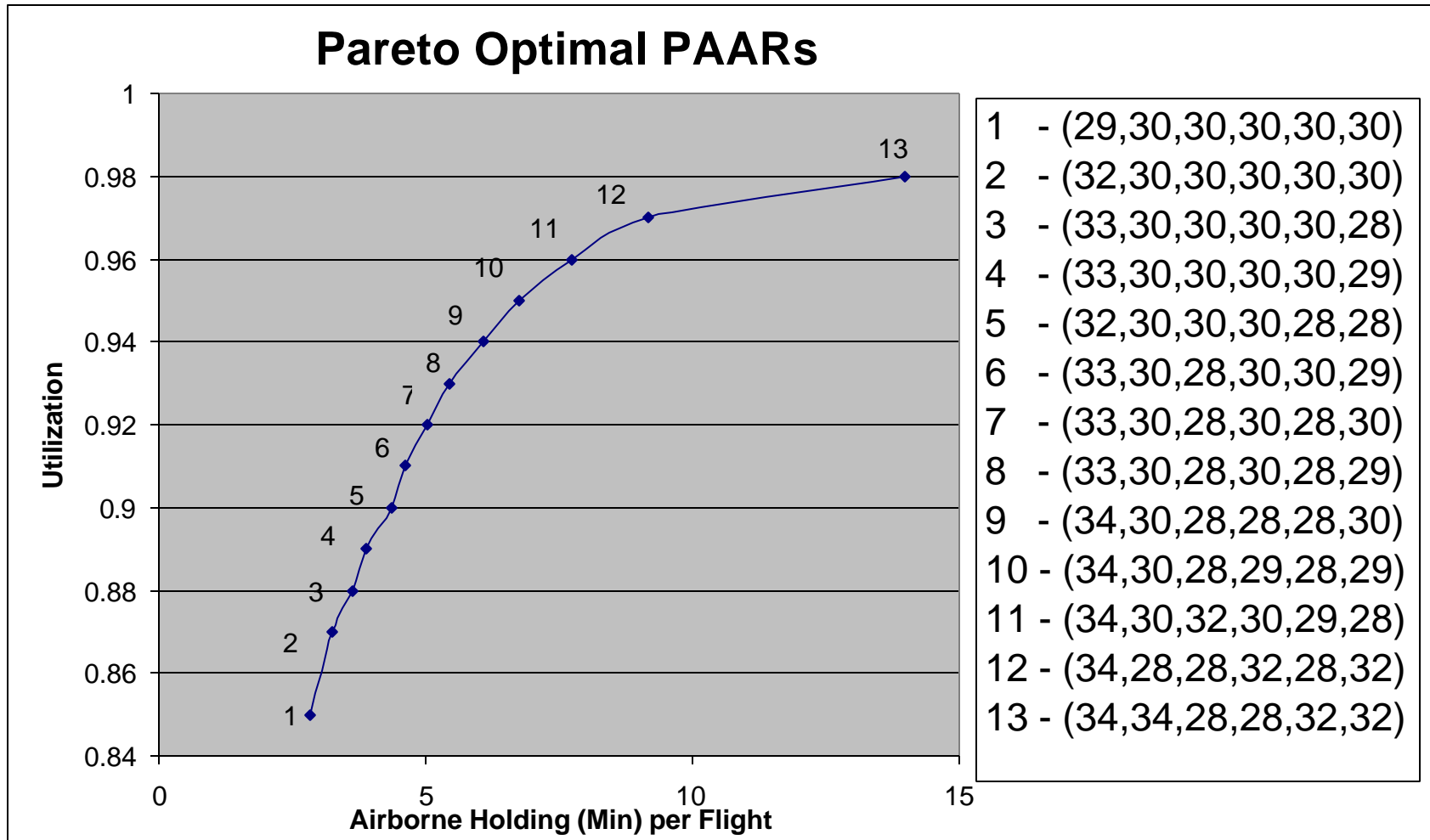
## ■ Pareto Optimality

A state  $A$  (a set of parameters) is said to be Pareto optimal, if there is no other state  $B$  dominating the state with respect to a set of objective functions.

A state  $A$  dominates a state  $B$ , if  $A$  is better than  $B$  in at least one objective function and not worse with respect to all other objective functions.

# Results for Simulation Model(contd.)

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# Summary

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- Significant stochasticity in airport arrival demand
- Demand Uncertainties lead to under-utilization, and excessive airborne holding
- Two models - SMIO and Simulation Model - are developed
- Models recommend policy changes in setting of PAARs - substituting *staggered patterns* for *flat patterns*