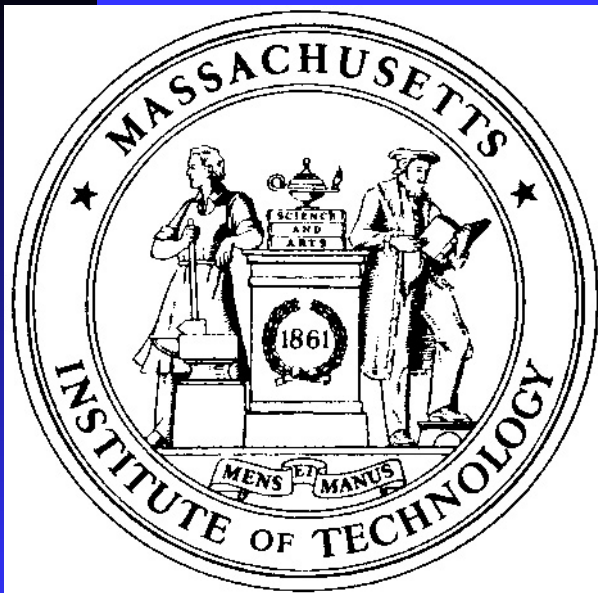


Service Network Design: Applications in Transportation and Logistics



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Problem Definition

- Service network design problems in transportation and logistics, subject to limited resources and variable service demands
 - ◆ Determine the cost minimizing or profit maximizing set of services and their schedules
 - ◆ What is the best location and size of terminals such that overall costs are minimized?
 - ◆ What is the best fleet composition and size such that service requirements are met and profits are maximized?

Service Network Design Applications

■ Examples

- ◆ Determining the set of flights and their schedules for an airline
- ◆ Determining the routing and scheduling of tractors and trailers in a trucking operation
- ◆ Jointly determining the aircraft flights, ground vehicle and package routes and schedules for time-sensitive package delivery

Research on Service Network Design

■ Rich history of research on network design applications

◆ *Network Design*

- ◆ Balakrishnan et al (1996); Desaulniers, et al (1994); Gendron, Crainic and Frangioni (1999); Gendron and Crainic (1995); Kim and Barnhart (1999); Magnanti (1981); Magnanti and Wong (1984); Minoux (1989)

◆ *Freight Transportation Service Network Design*

- ◆ Armacost, Barnhart and Ware (2002); Crainic and Rousseau (1986); Crainic (2000); Farvolden and Powell (1994); Lamar, Sheffi and Powell (1990); Newton (1996); Ziarati, et al (1995)

◆ *Fleet Routing and Scheduling*

- ◆ Appelgren (1969, 1971); Desaulniers, et al (1997); Desrosiers, et al (1995); Dumas, Desrosiers, Soumis (1991); Leung, Magnanti and Singhal (1990); Ribeiro and Soumis (1994)

Challenges

- Service network design problems in transportation and logistics are characterized by
 - ◆ Costly resources, tightly constrained
 - ◆ Many highly interdependent decisions

Both models and algorithms are critical to tractability

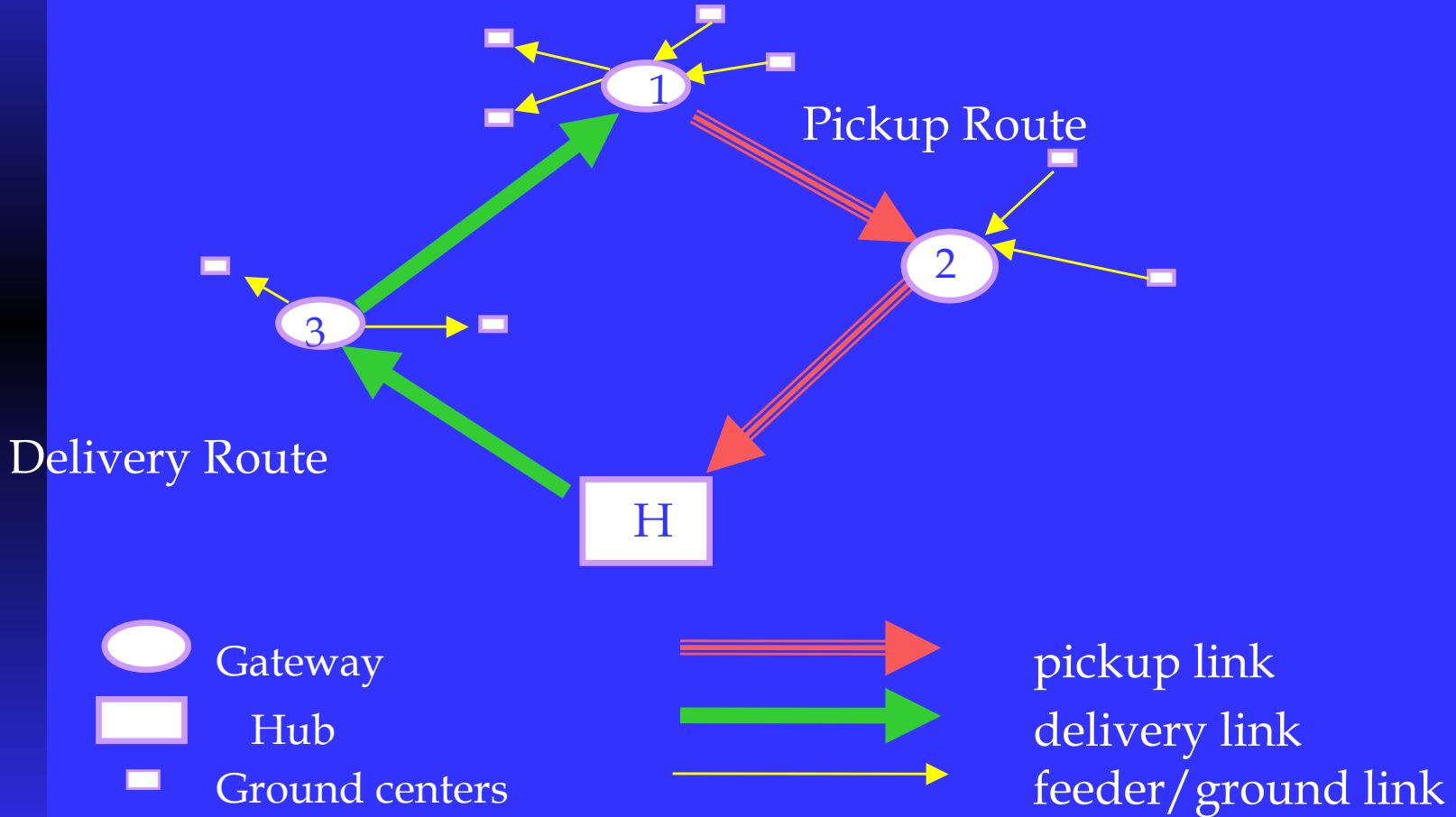
- ◆ Associated with sets of design decisions, not a single design decision
- *Huge* mathematical programs
- Notoriously weak linear programming relaxations

Designing Service Networks for Time-Definite Parcel Delivery

- Problem Description and Background
- Designing the Air Network
 - ◆ Optimization-based approach
- Case Study

Research conducted jointly with
Prof. Andrew Armacost, USAFA

Problem Overview



UPS Air Network Overview

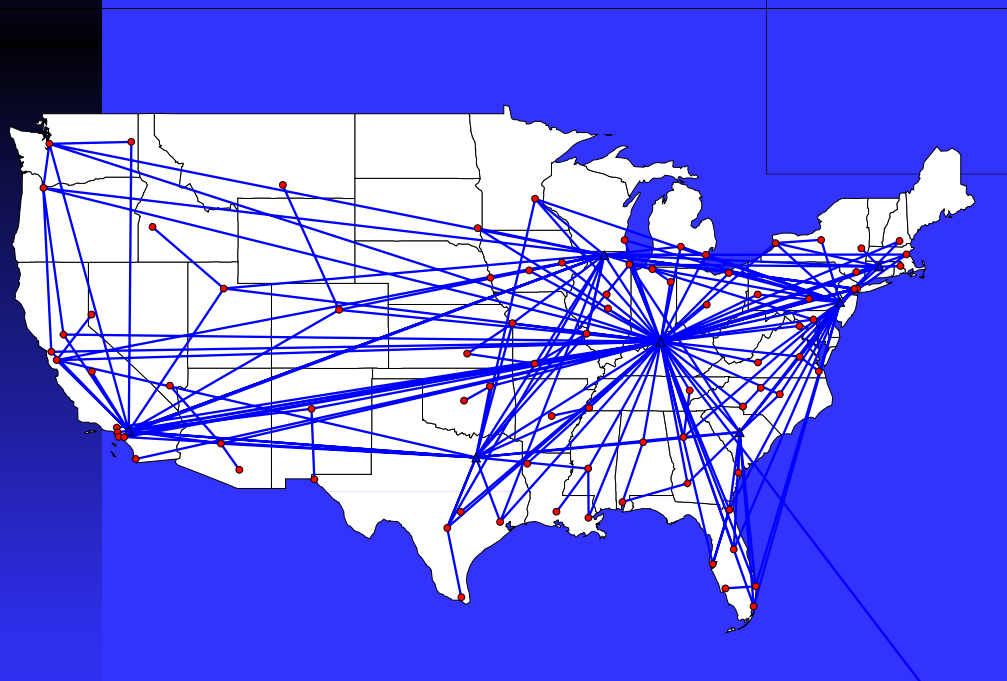
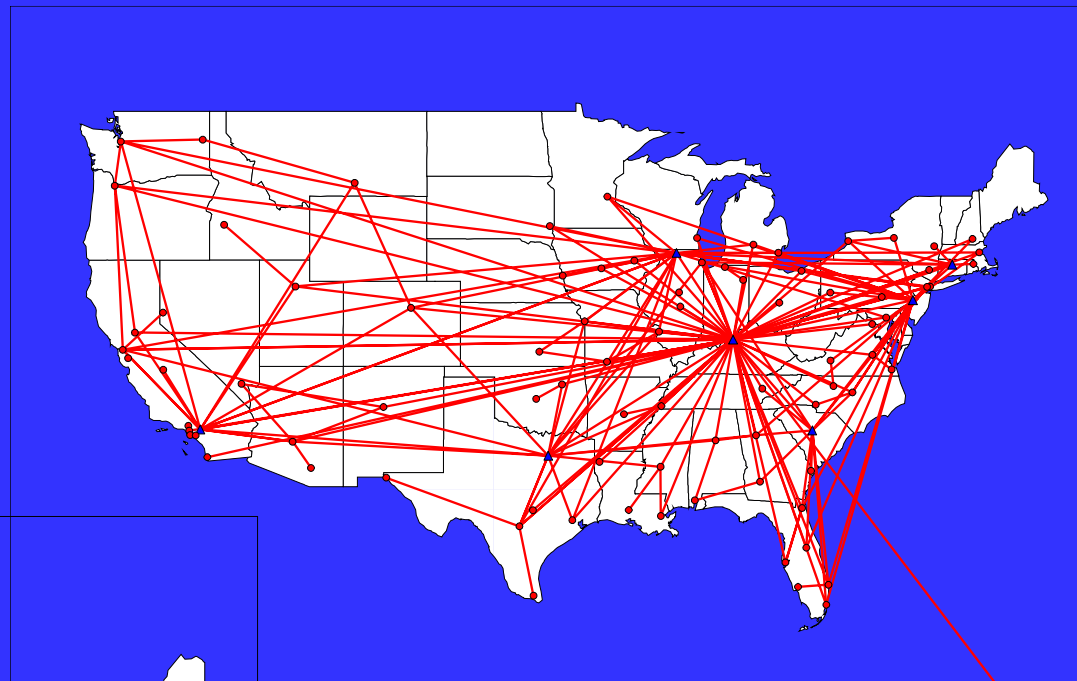
- Aircraft
 - ◆ 168 available for Next-Day Air operations
 - ◆ 727, 747, 757, 767, DC8, A300
- 101 domestic air “gateways”
- 7 hubs (Ontario, DFW, Rockford, Louisville, Columbia, Philadelphia, Hartford)
- Over one million packages nightly

Research Question

- What aircraft routes and schedules provide on-time service for all packages while minimizing total costs?

UPS Air Network Overview

Delivery Routes



Pickup Routes

Problem Formulation

- Select the minimum cost routes, fleet assignments, and package flows
- Subject to:
 - ◆ Fleet size restrictions
 - ◆ Landing restrictions
 - ◆ Hub sort capacities
 - ◆ Aircraft capacities
 - ◆ Aircraft balance at all locations
 - ◆ Pickup and delivery time requirements

Express Shipment Service

Model 1: Dijkstra's Problem

$$\text{min} \sum_{k \in K} \sum_{(i,j) \in A} c_{ij}^k x_{ij}^k + \sum_{f \in F} \sum_{r \in R^f} d_r^f y_r^f$$

$$\sum_{k \in K} x_{ij}^k \leq \sum_{f \in F} \sum_{r \in R^f} \delta_{ij}^{f,r} u_r^f \quad (i,j) \in A$$

$$\sum_{j:(i,j) \in A} x_{ij}^k - \sum_{(j,i) \in A} x_{ji}^k = \begin{cases} b^k & \text{if } i = O(k) \\ -b^k & \text{if } i = D(k) \end{cases}$$

$$\sum_{k \in K} \delta_{ik}^h \leq a_i \quad h \in H$$

$$\sum_{r \in R^f} \beta_i^r y_r^f = 0 \quad i \in N, f \in F$$

$$\sum_{r \in R^f} u_r^f \leq n^f \quad f \in F$$

$$\sum_{f \in F} \sum_{r \in R^f} \delta_{ij}^{f,r} u_r^f \leq a_{ij} \quad h \in H$$

$$x_{ij}^k \geq 0 \quad (i,j) \in A, k \in K$$

$$y_r^f \in \mathbb{Z}_+ \quad r \in R^f, f \in F$$

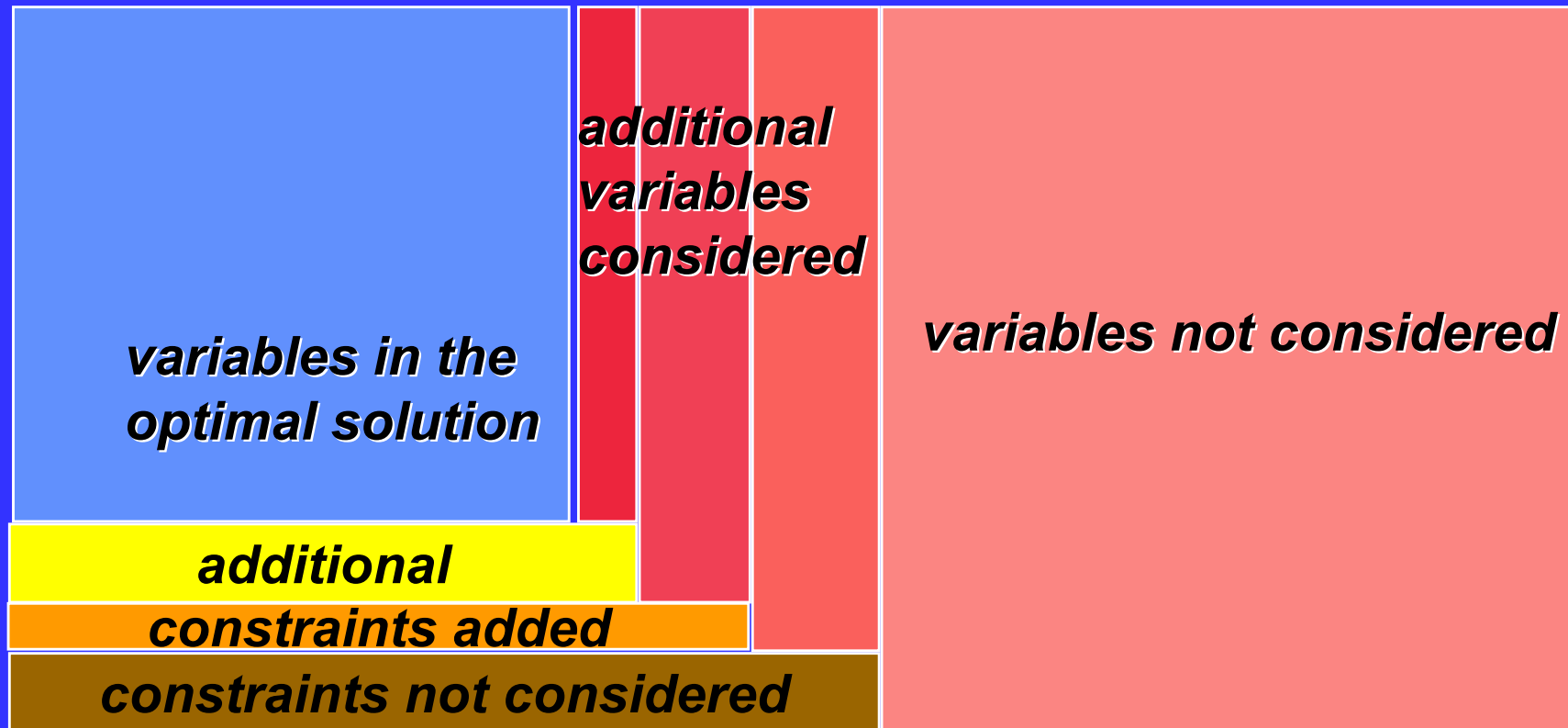
The Size Challenge

- Conventional model
 - ◆ Number of variables exceeds one billion
 - ◆ Number of constraints exceeds 200,000

Column and Cut Generation

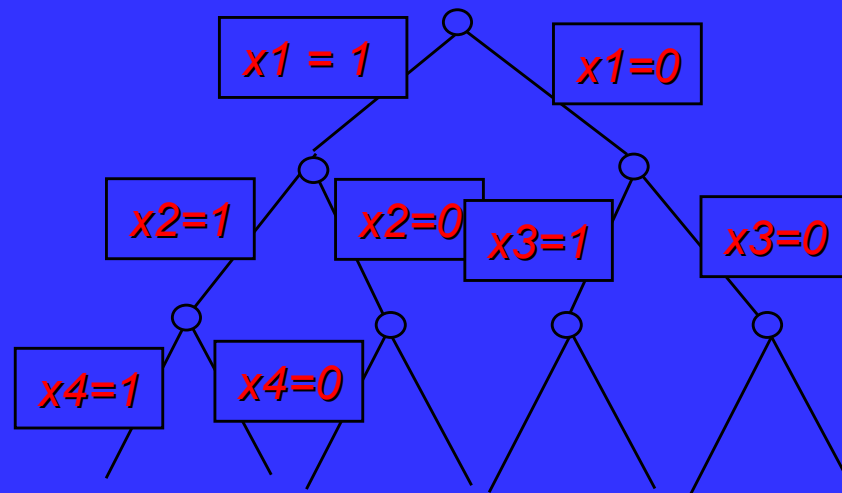
Constraint Matrix

billions of variables



Algorithms for Huge Integer Programs: Branch-and-Price-and-Cut

- Determines Optimal Solutions to Huge Integer Programs
 - ◆ Combines Branch-and-Bound with Column Generation and Cut Generation to solve the LP's



The Integrality Challenge

- Initial feasible solution about triple the best bound
 - ◆ Multiple day runtimes to achieve first feasible solution

Resolution of Challenges

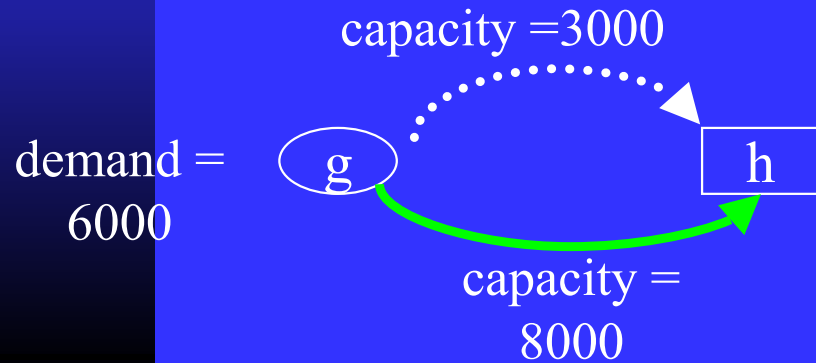
- Algorithms are not enough
- Key to successful solution of these very large-scale problems are the models themselves

Alternative Formulations

- A given problem may have different formulations that are all logically equivalent yet differ significantly from a computational point of view
- This has motivated the study of systematic procedures for generating and solving *alternative* formulations

Reformulation: Key Ideas

Aircraft Route Variables



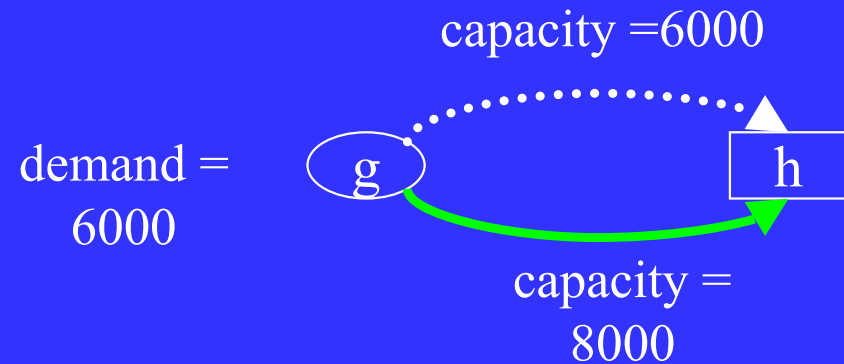
Capacity-demand:

$$3000y_1 + 8000y_2 \geq 6000$$

Cover:

$$y_1 + y_2 \geq 1$$

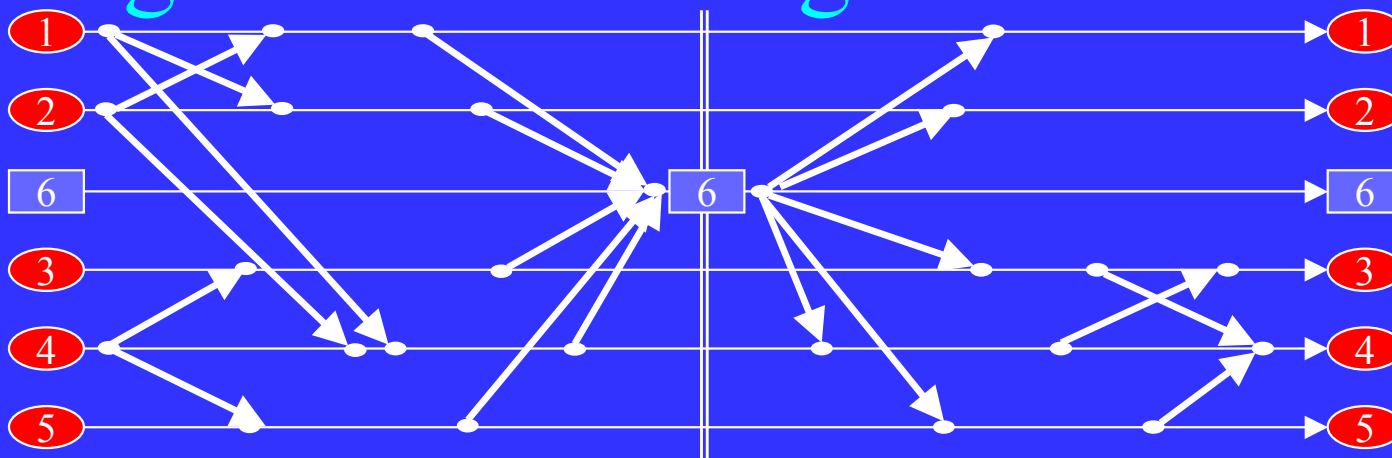
Composite Variables



Cover:

$$y_3 + y_2 \geq 1$$

Strength Results: Single Hub Example



Time-Space Representation of
Plane/Package Movements

	ESSND	ARM
Rows	53	34
Cols	67	42
NZ	274	255
LP Solution	10663	28474
IP Solution	28474	28474
B&B Nodes	781	1
LP-IP Gap	167%	0%

ARM vs UPS Planners

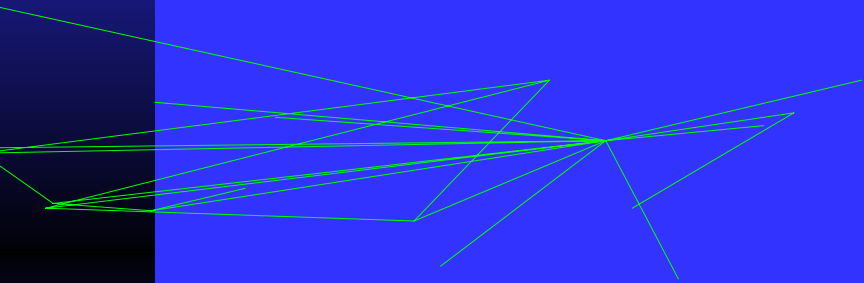
Minimizing Operating Cost for UPS

- Improvement (reduction)
 - ◆ Operating cost: 6.96 %
 - ◆ Number of Aircraft: 10.74 %
 - ◆ Aircraft ownership cost: 29.24 %
 - ◆ Total Cost: 24.45 %
- Running time
 - ◆ Under 2 hours

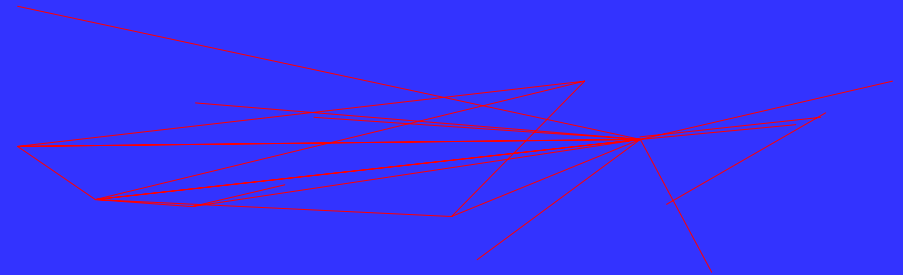
ARM vs. Planners

Routes for One Fleet Type

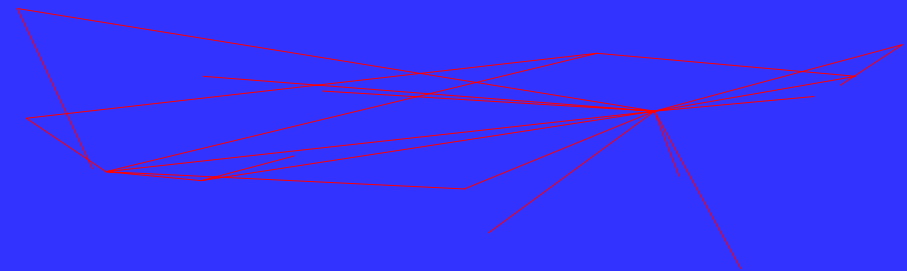
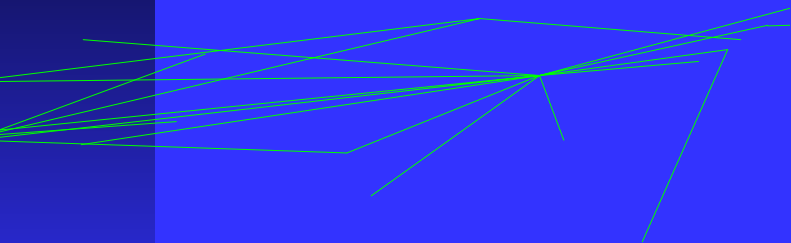
Pickup Routes



Delivery Routes



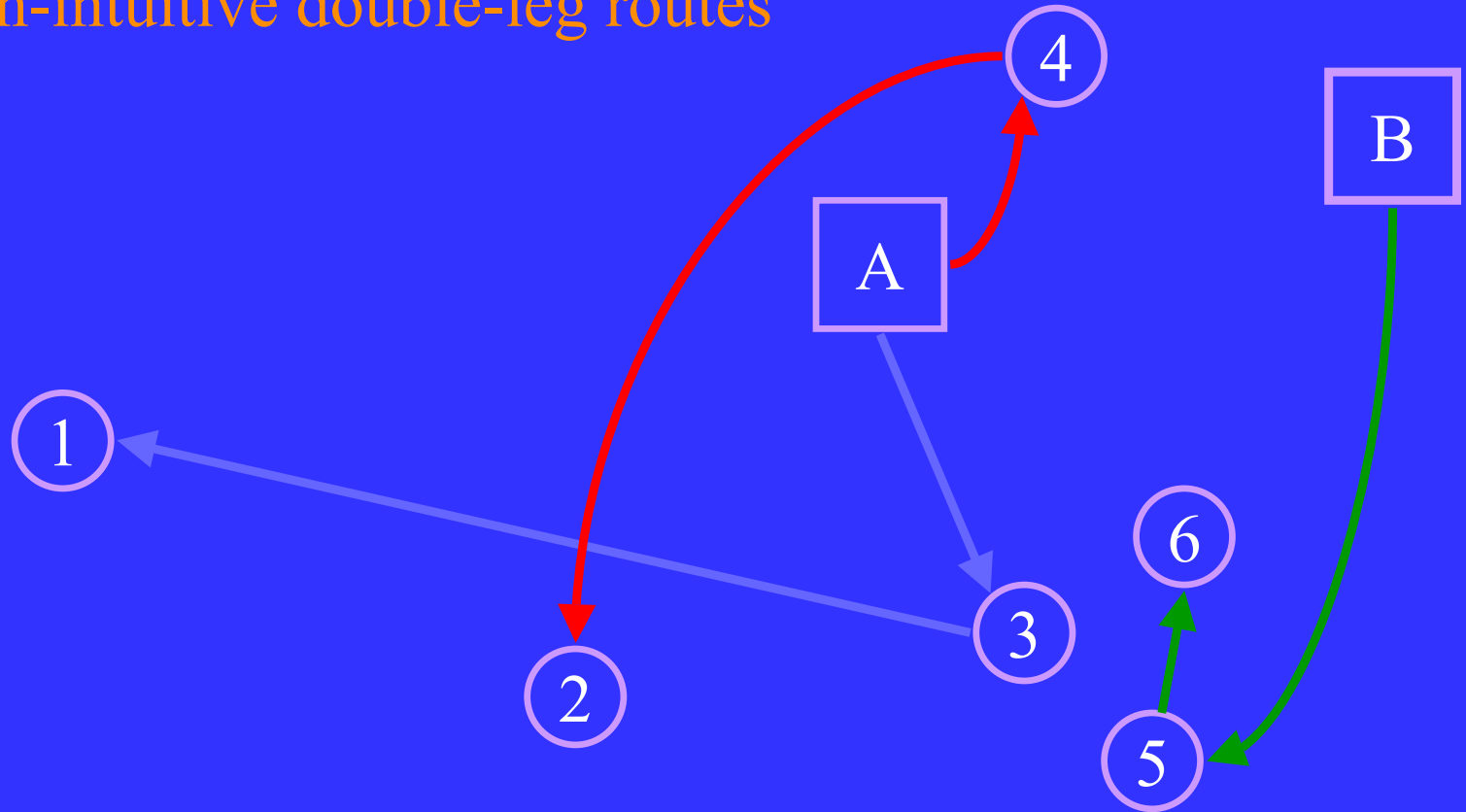
Planners' Solution



ARM Solution

ARM Solution

Non-intuitive double-leg routes



Model takes advantage of timing requirements, especially in case of A-3-1, which exploits time zone changes

Model takes advantage of ramp transfers at gateways 4 and 5

Conclusions

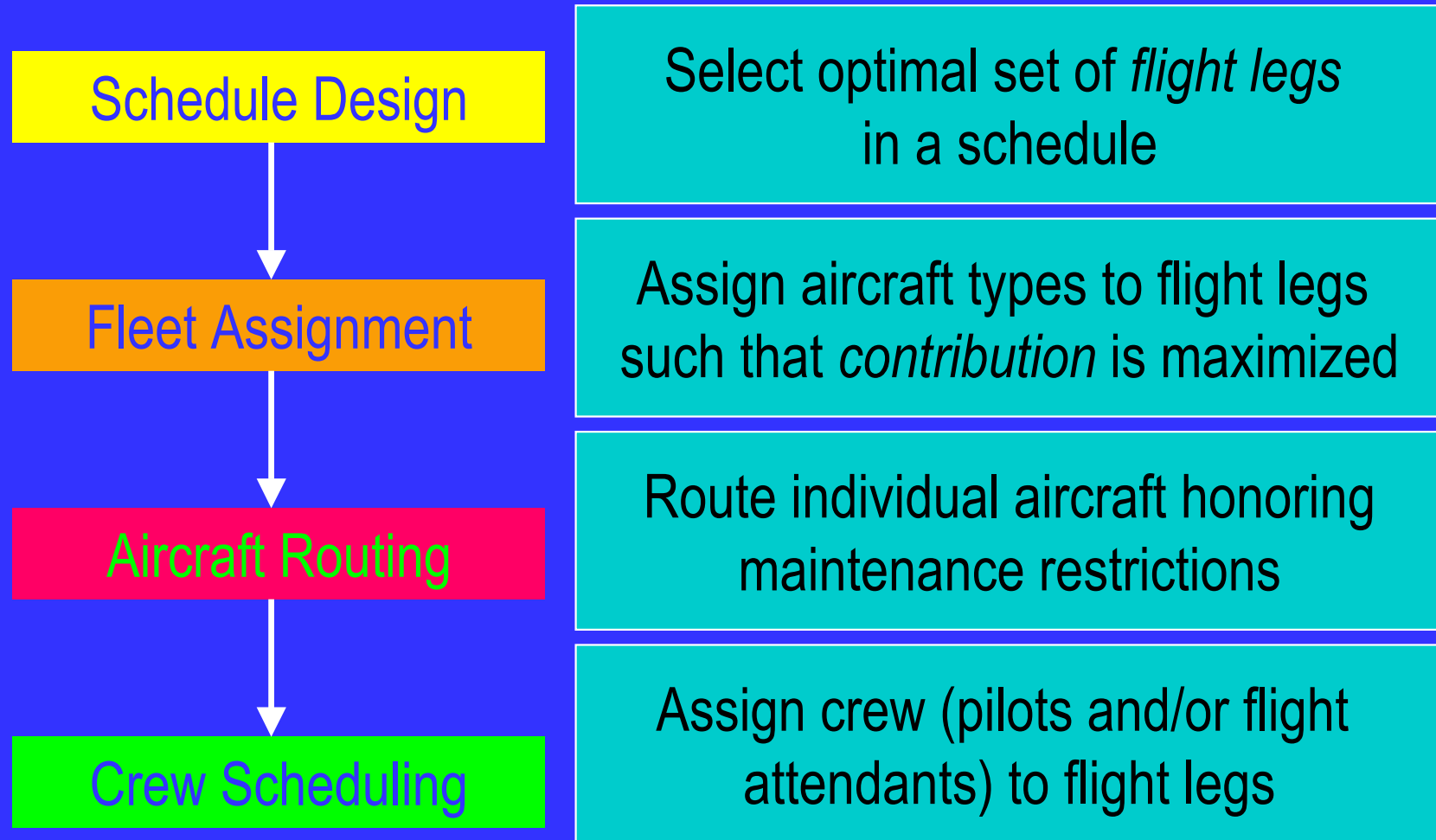
- Solving large-scale service network design problems
 - ◆ Blend art and science
 - ◆ Composite variable modeling can often facilitate
 - ◆ Tractability
 - ◆ Extendibility

Service Network Design and Passenger Service in the Airline Industry

- Problem Description and Background
- Analysis
- Some Research Findings

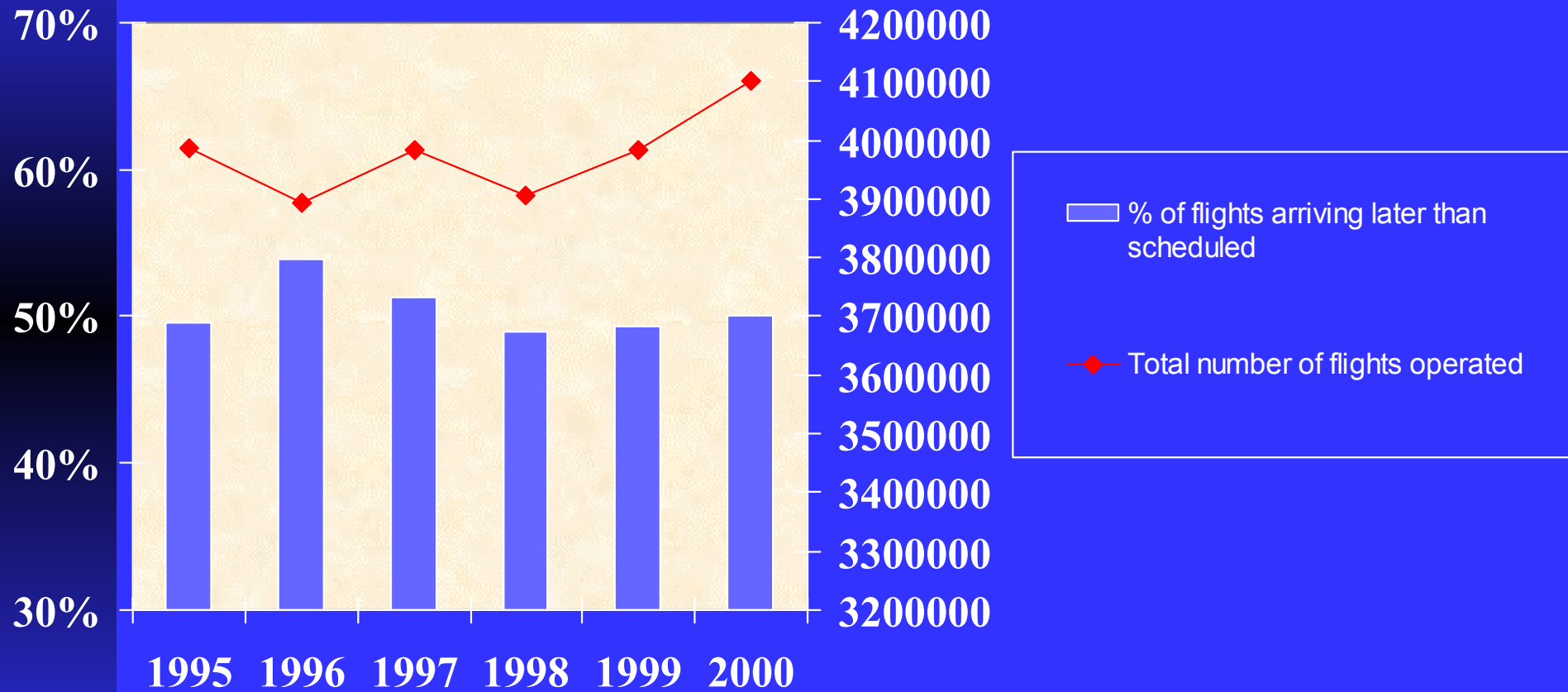
Research conducted jointly with
Stephane Bratu

Airline Schedule Planning

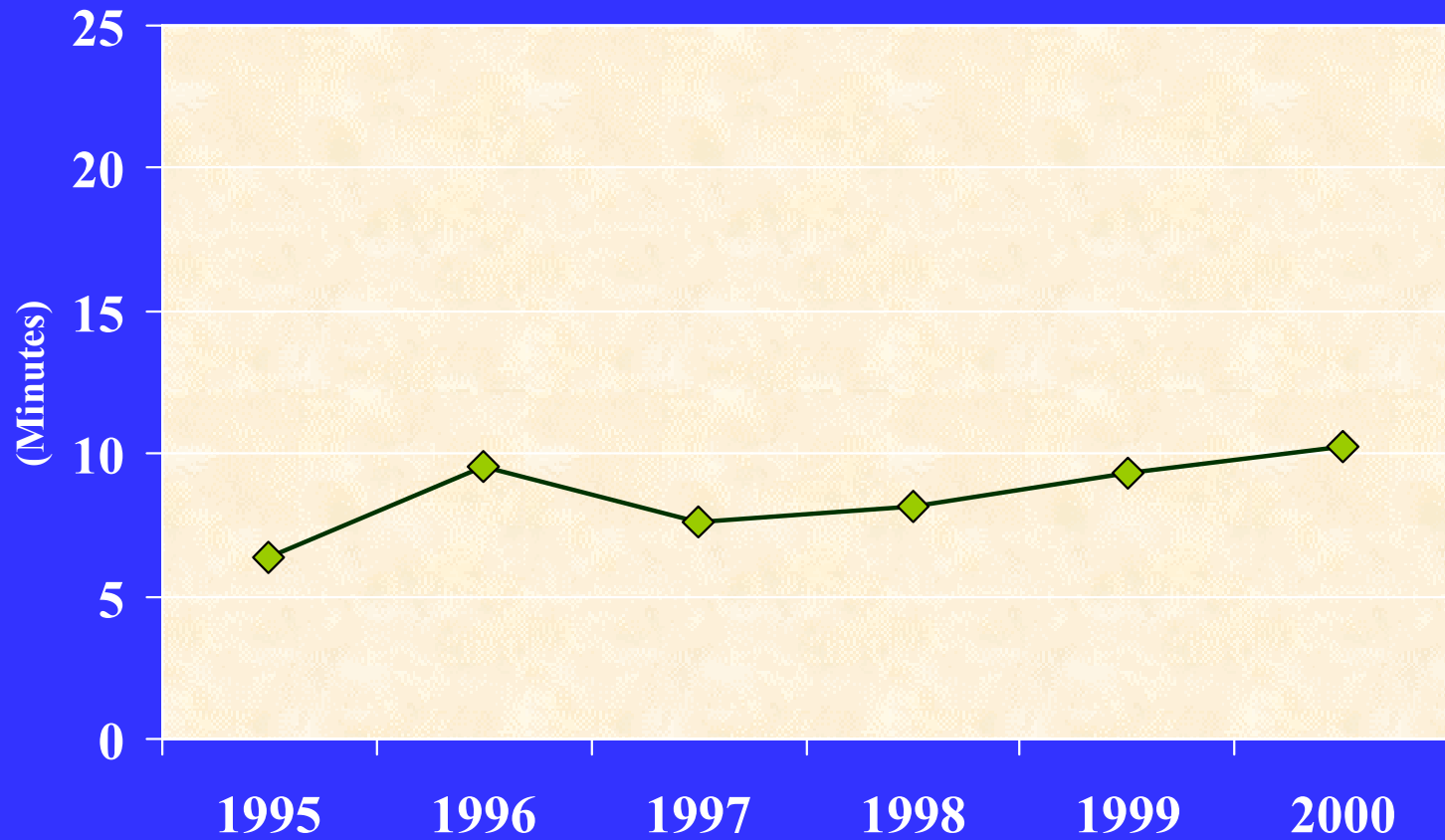


Some Simple Statistics ...

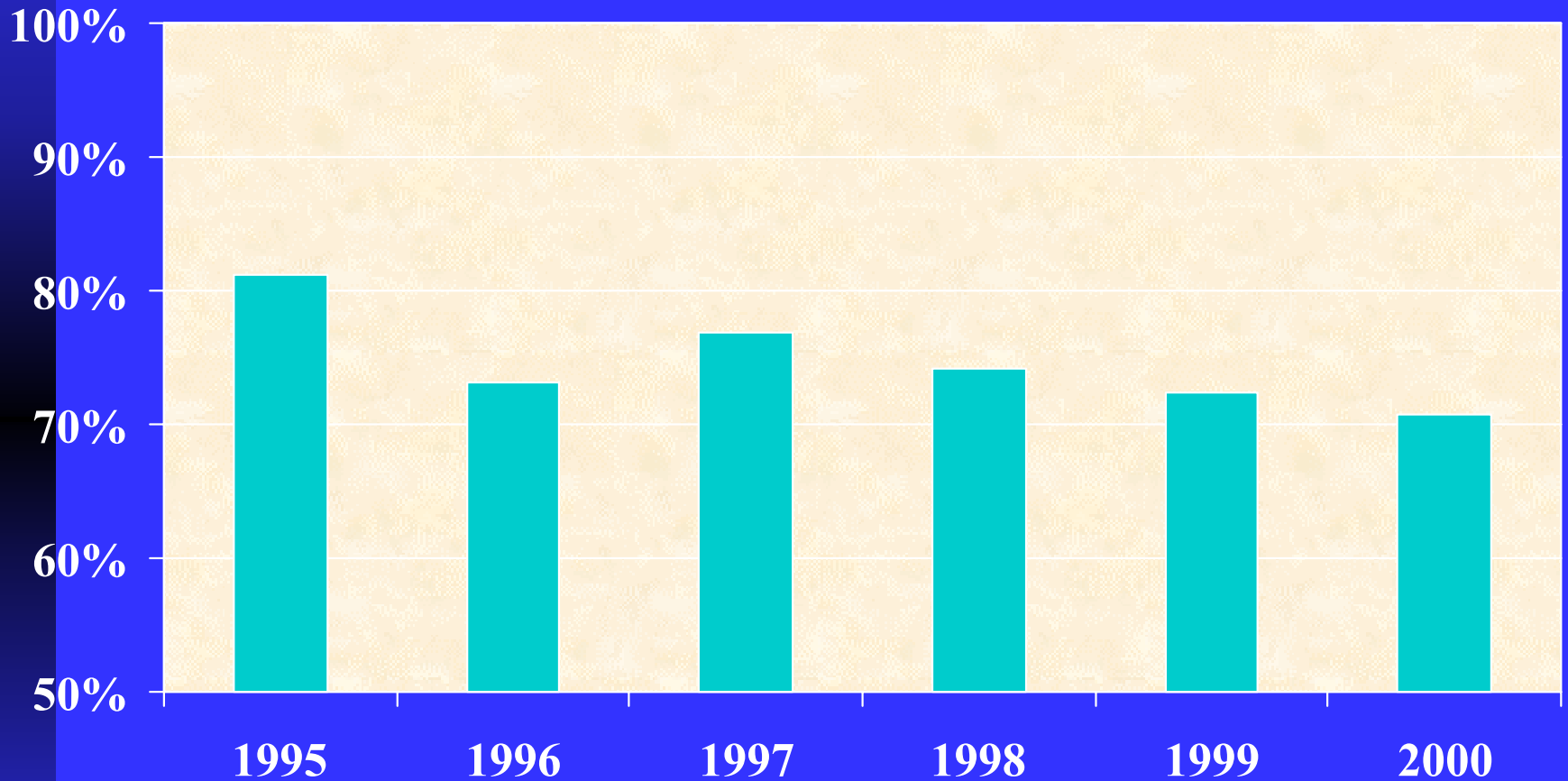
Number and Percentage of Delayed Flights



Average Delay Duration of Operated Flights



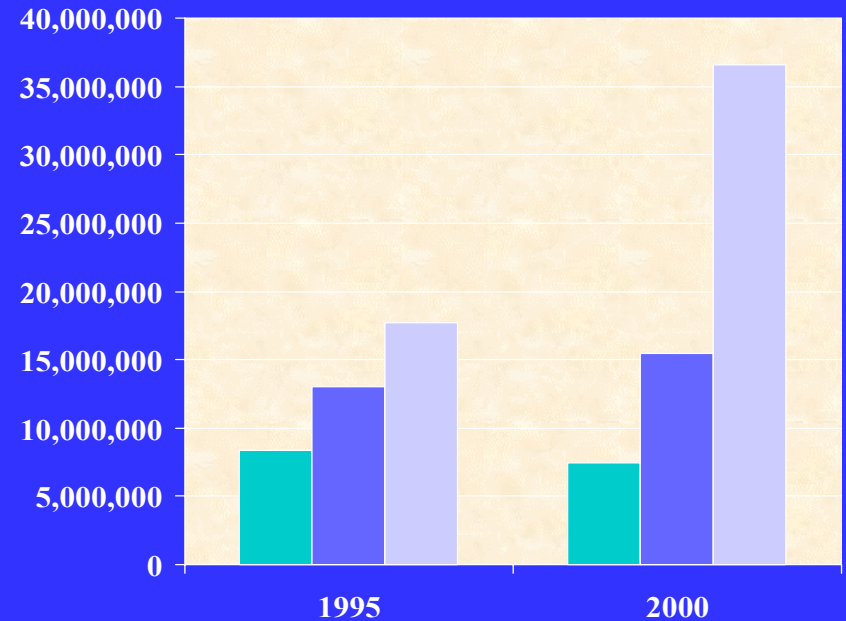
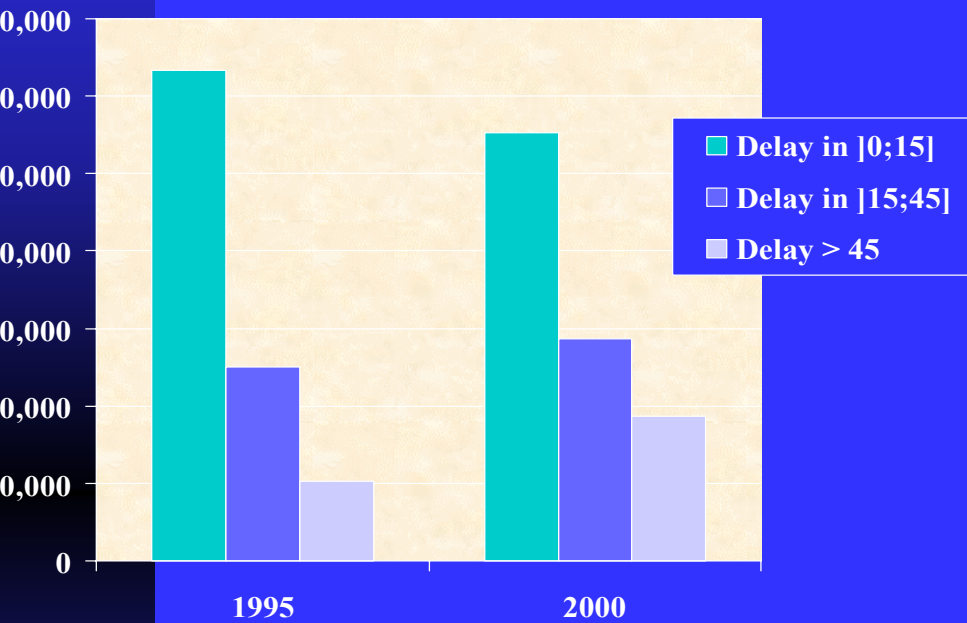
15-minute On-Time Performance



Important Factors Not Accounted For in Simple Statistics

1. Distribution of flight delays
2. Hub-and-Spoke Networks
3. Flight cancellation rate

Factor 1: Shift to Longer Flight Delays

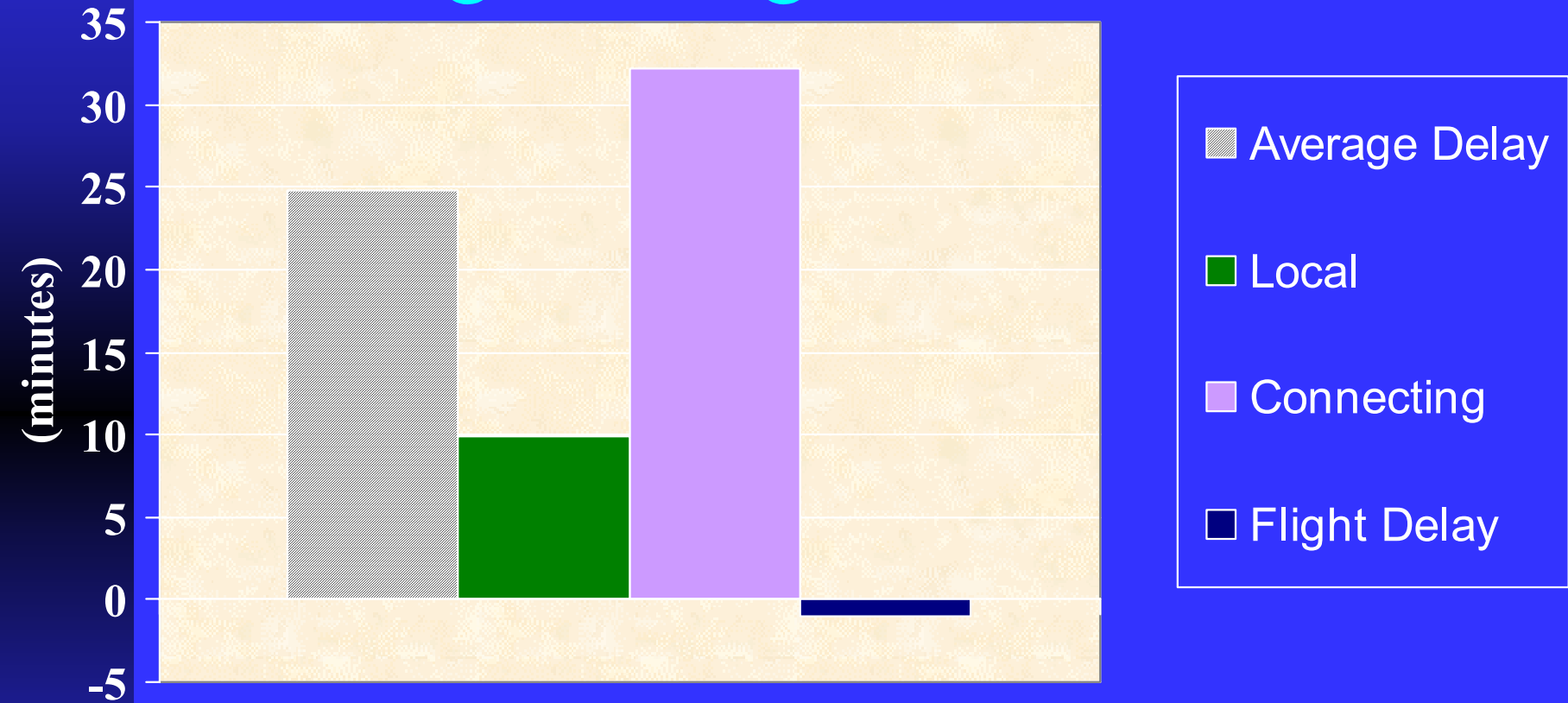


Number of Delayed Flights

Total Delay Minutes

The delay distribution has shifted from short to long delays

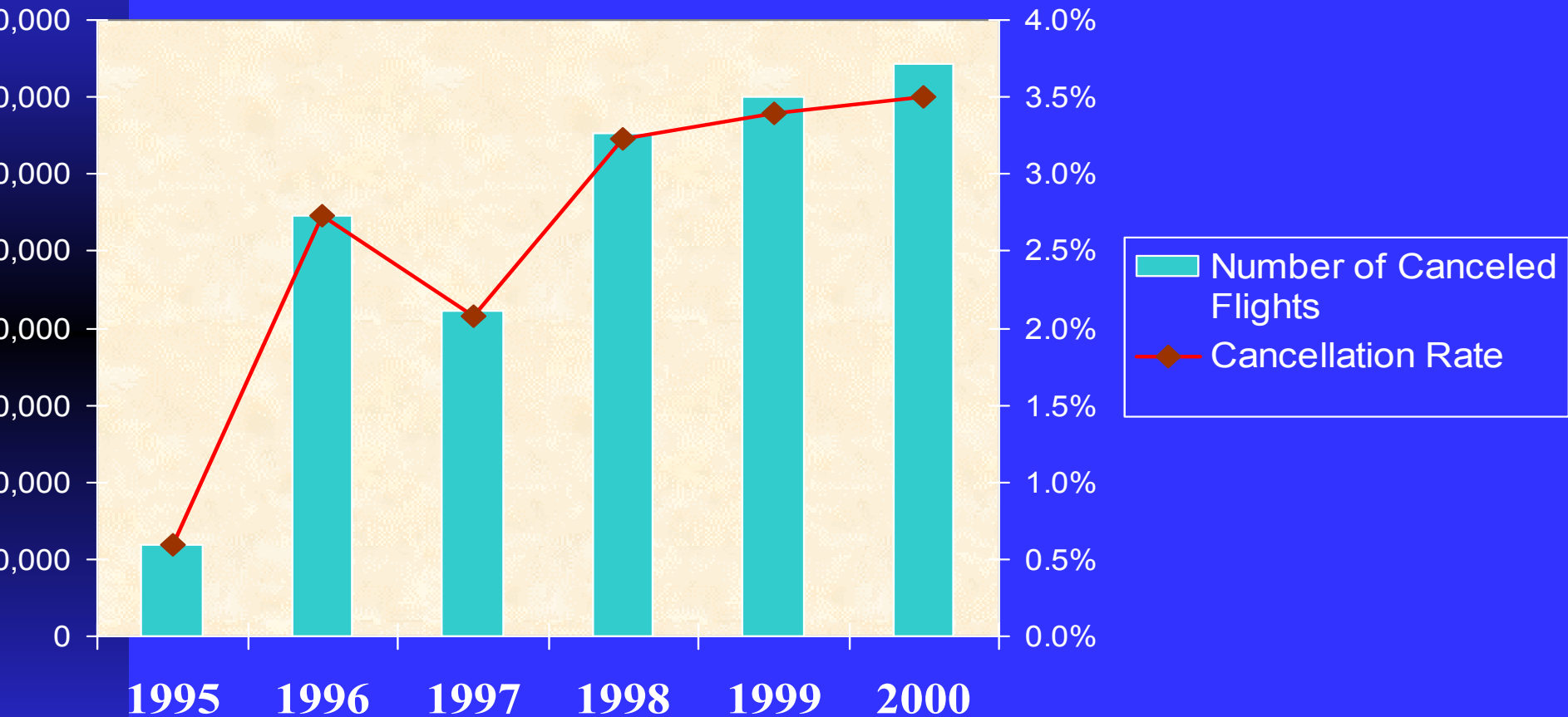
Factor 2: Hub-and-Spoke and Connecting Passengers



Average Passenger and Flight Delays

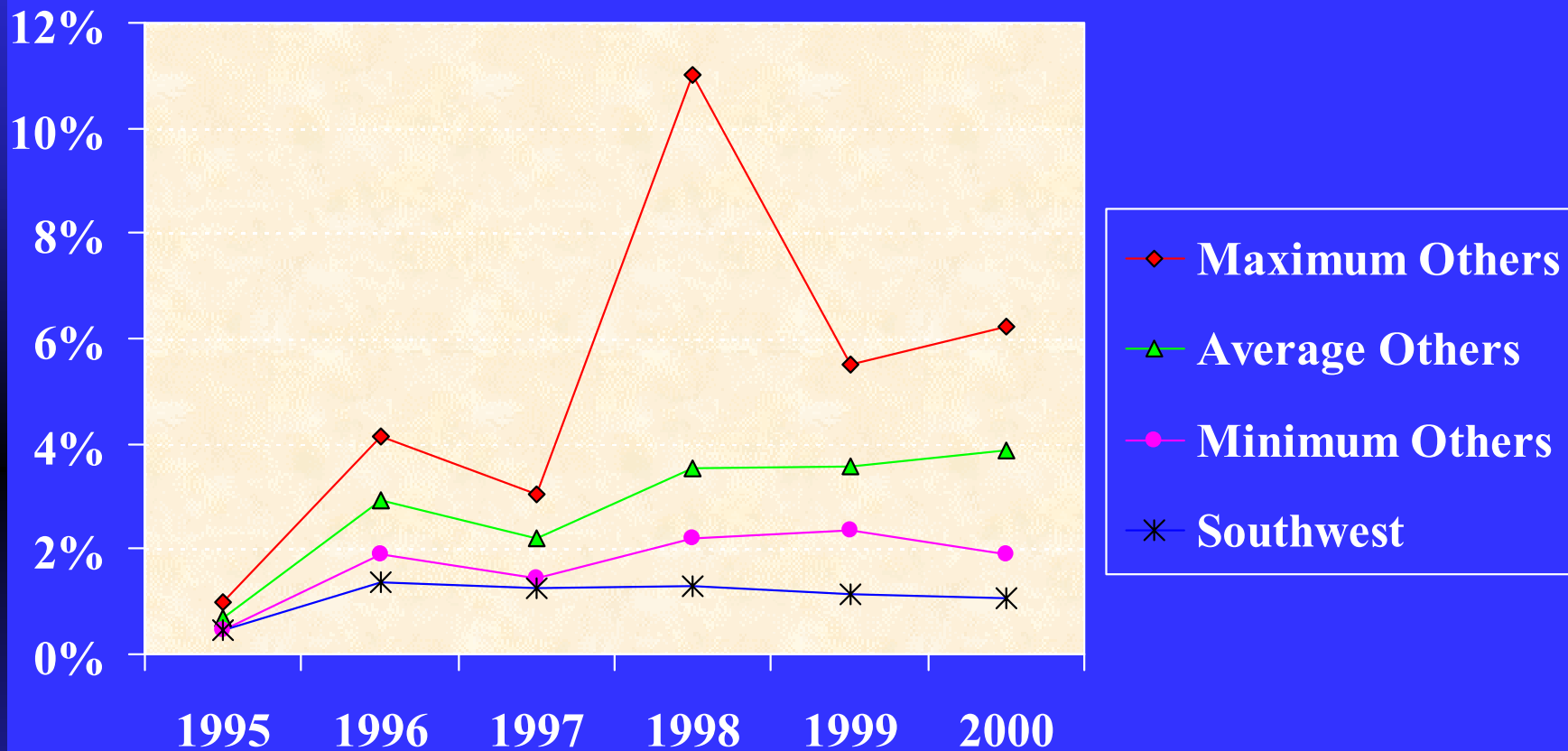
Flight delays underestimate passenger delays
Key explanation lies in the connecting passengers

Factor 3: Number of Canceled Flights and Cancellation Rates



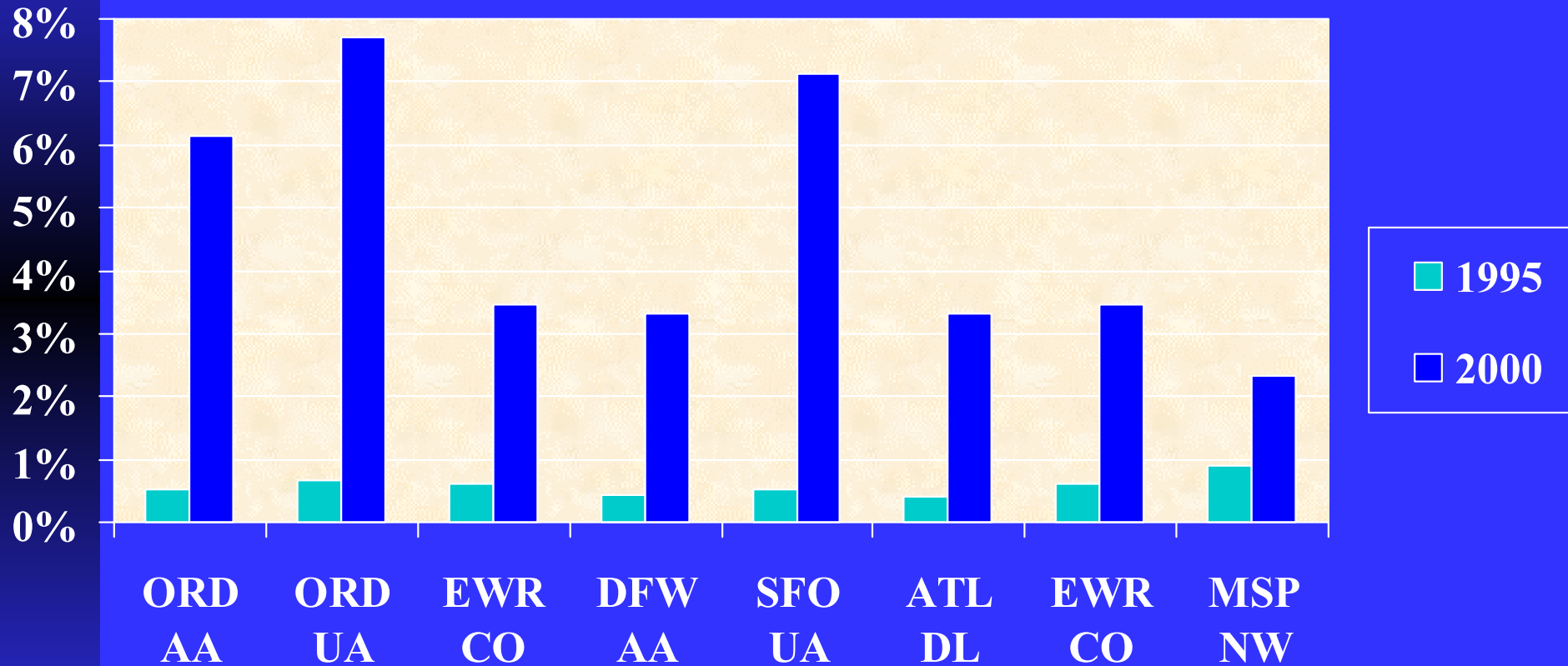
Delay statistics do not consider cancellations

Cancellation Rate: Southwest and the Other Majors Airlines



Southwest has a lower cancellation rate than any other Major from 1995 to 2000 due in part to increases in cancellation rates at some congested hubs

Hub Cancellation Rates



Network Design and Passenger Service

- DOT 15 minute on-time-performance is inadequate
- There are a number of alternative “flight schedules” with similar associated costs and profitability, but vastly different associated passenger delays
 - ◆ *Service* network design needs to incorporate *service considerations*
- Flight cancellations can reduce overall passenger delay
 - ◆ High load factors together with flight delays can result in excessive passenger delays
- De-banking can result in much longer planned connection times, but only slightly longer connection times in actuality