

Management of Bus Operations and Technology

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June 10th, 2008

OUTLINE

- **Applications of Automated Data Collection Systems**
 - **Scheduling**
 - **Connection Protection**
 - **Service Quality Monitoring**
 - **Travel Pattern Inference**
 - **Travel Behavior**
- **Ongoing Bus Operations Research**
 - **Scheduling**
 - **Limited Stop / X Service Design**
 - **Operations Management and Control**
 - **Interchange Analysis**

MIT Research Focus

- **Get better information from Automated Data Collection Systems**
 - **AFC: Automatic Fare Collection Systems**
 - **AVL: Automatic Vehicle Location Systems**
 - **APC: Automatic Passenger Counting Systems**
- **To support key agency functions:**
 - **Service and Operations Planning**
 - **Operations Management and Control**
 - **Customer Information**
 - **Performance Measurement and Monitoring**

Scheduling Application

Context: Rail scheduling and operations control

Data: Train tracking data

Focus: Analysis of train dwell times, headways, schedule adherence, time-space diagrams, animation playback capability

Application: MBTA Red Line

Recommendations:

Schedule adjustments, branch offsets adjustments, tighter terminal departure discipline

Key researchers: Haris Koutsopoulos, Matt Dixon, Zhigao Wang (2006)

Connection Protection Application

Context: Transfers from rail to bus

Data: Train tracking data

Focus: Develop improved but simple dispatching strategy for buses based on bus holding lights linked to impending train arrival after long gap

Application: MBTA Red Line Alewife Station

Recommendations:

Implementation would reduce transfer wait time by 25% and greatly reduce "near misses"

Key researchers: Drew Desautels (2006)

Service Quality Monitoring Application

Context: Rail operations

Data: Smart card station entry and exit times

Focus: Develop measures of service reliability from customer's perspective

Application: London Underground service times using Oyster time data

Recommendations:

Use of reliability buffer as an additional measure of service delivery alongside mean measure (Journey Time Metric)

Key researchers: Joanne Chan (2007), David Uniman (in progress)

Travel Pattern Inference Application

Context: Any public transport network -- bus, rail, or combined

Data: AFC transactions (smart card or magnetic stripe) and AVL data

Focus: Estimation of customer origin-destination travel patterns

Application: CTA Rail network
London Underground network
CTA Bus network

Recommendations:

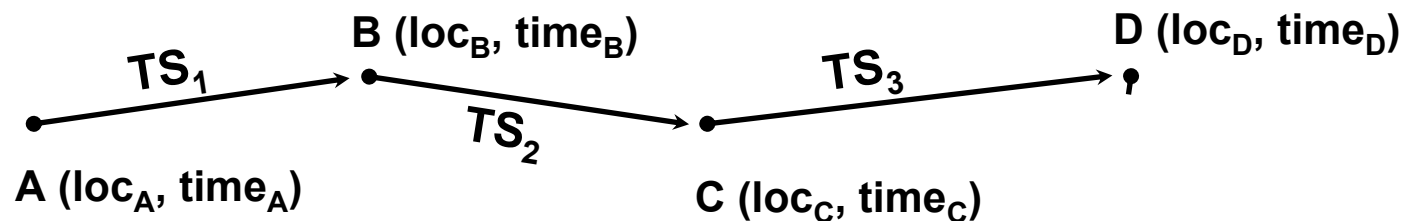
A practical method for estimating travel patterns with virtually no additional cost beyond existing ADCs

Key researchers: Jinhua Zhao (2004), Alex Cui (2006), Fabio Gordillo (2006), Joanne Chan (2007)

Basic Idea

Each AFC record includes:

- AFC card ID
- transaction type
- transaction time
- transaction location: rail station or bus route



The destination of many trip segments (TS) is also the origin of the following trip segment.

- Note:**
- 1) each bus boarding requires a new AFC transaction:
TS_{bus} represents an unlinked bus trip
 - 2) rail-rail transfers do not require a new AFC transaction:
TS_{rail} represents a path on the rail network

Travel Behavior Application

Context: Any public transport network -- bus, rail, or combined

Data: Registered smart card address (exact or approximate), smart card transactions, and AVL data

Focus: Infer public transport access behavior and modal preferences

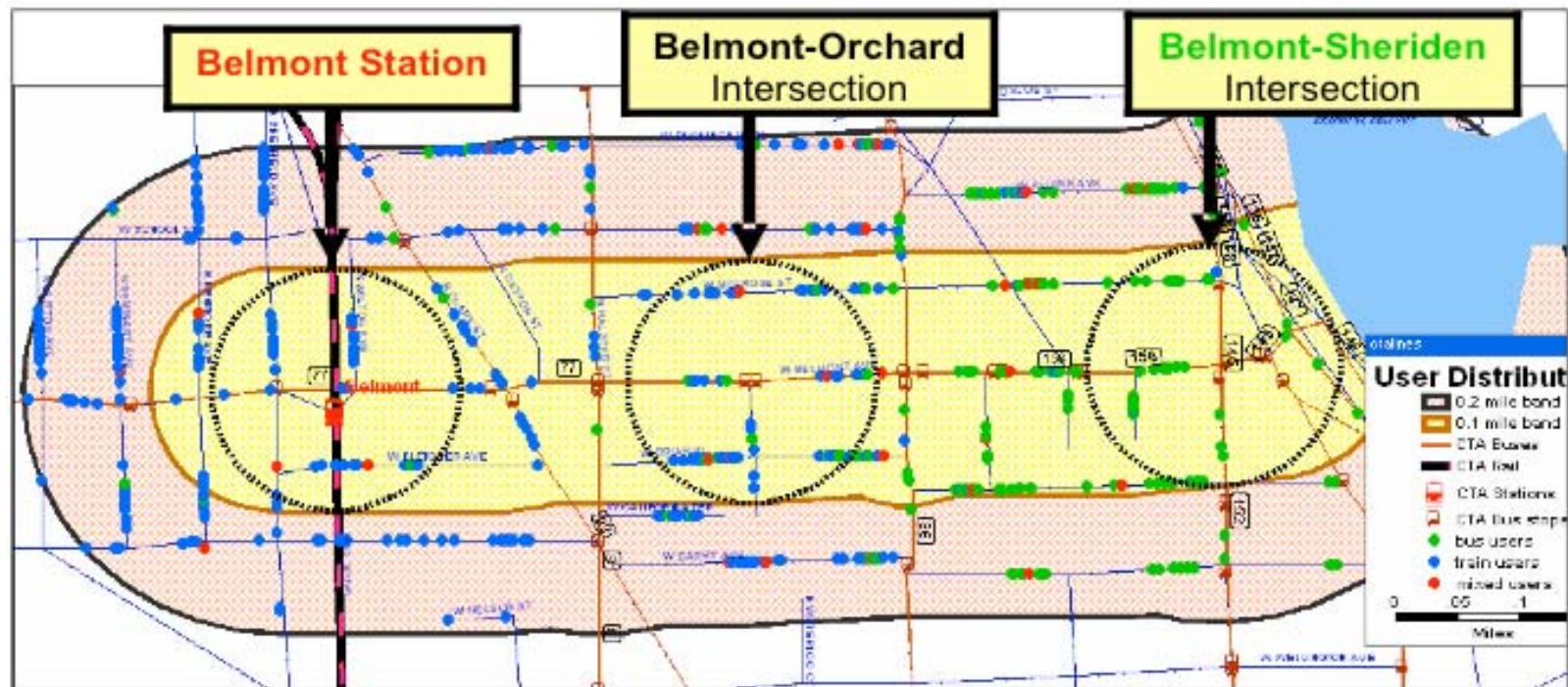
Application: CTA bus and rail network

Recommendations:

This represents a valuable way to monitor behavior and modal preferences as the system changes over time

Key researchers: Mariko Utsunomiya (2005), Saumya Gupta (2006)

Path Choice Analysis: Sample Users



- Multiple rail and bus routes serving the loop
- High quality express bus service and rail service

Ongoing Bus Operations Research

- **Scheduling**
- **Limited Stop / X Service Design**
- **Operations Management and Control**
- **Interchange Analysis**
- **Bus Route Simulation Model Development**
- **Chicago Loop Congestion Analysis**

Bus Scheduling

- **Bus service reliability is a chronic concern of customers and agencies**
- **Traditionally schedules are developed using rules of thumb -- not a real problem because data has been very poor on actual running times**
- **Now with AVL data we can evaluate alternative scheduling methods**
- **Goal is to develop schedules which:**
 - **result in reliable service**
 - **don't increase running time too much**
 - **don't cost too much**

Key researchers: Grace Fattouche (2007), Clara Yuan (in progress)

Problem Statement

On high-frequency service, i.e. headways ≤ 10 -12 minutes:

- **Develop a timepoint level schedule which minimizes the total weighted time cost to customers**

Subject to:

- **At least 90% of buses should be ready to start their next trip on time**
- **Operators should have at least 5 minutes of recovery time at one end of the route**

Model

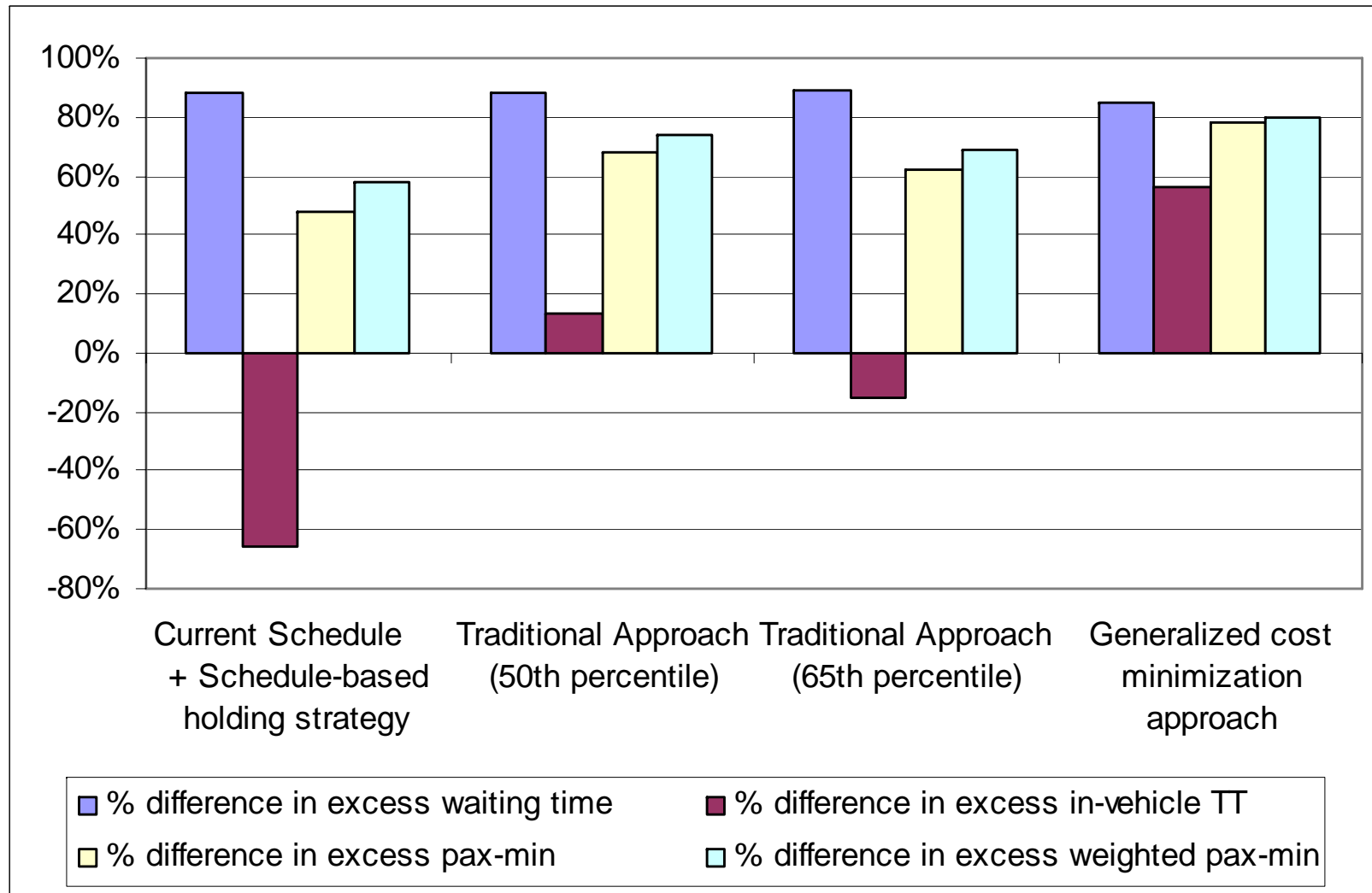
- **Evaluates the cost for the waiting passengers, onboard passengers and CTA of a proposed schedule**
- **Assumes schedule-based holding strategy (i.e. operators do not depart time point early)**
- **Inputs: AVL and APC data, headway, segment running times and number of buses in operations**
- **Outputs: costs for the waiting passengers, onboard passengers and CTA**



Application to CTA Route 95E

- CTA Key route
- Runs 5 miles East-West on 93rd St. and 95th St.
- Five time points
- Connects with the southern Red Line terminal, and two Metra stations
- High frequency: every 10 mins between 6:00 and 18:00

Percentage Change in Passenger Excess Time



Sensitivity Analysis

The generalized cost minimization schedule is sensitive to:

- **Ratio of waiting passengers to through passengers**
- **Location of the segment on the route**
- **Ratio of waiting passengers on later segments to through passengers**
- **Route Length**

Limited Stop / X Service Design

- **Many high-frequency bus routes in major metropolitan areas are long and have many stops, resulting in:**
 - long travel times
 - poor reliability
 - unattractive for long journeys
- **Limited Stop / X Services are overlay routes with far fewer stops, which:**
 - provide alternative service mix, which attracts different markets, e.g. longer journeys
 - reduce the interaction between buses, i.e., less bunching
- **One step towards BRT**

Key researchers: Stacey Schwarcz (2004), Harvey Scorcio (in progress)

Problem Statement

- **Establish guidelines for the addition of limited stop service**
- **Develop a model to help in design and evaluation of these services**

Key elements in limited stop service design:

- **Reduction in # of stops**
- **Running time savings**
- **Headway split between local and limited stop service**
- **Resources: unchanged or increased**

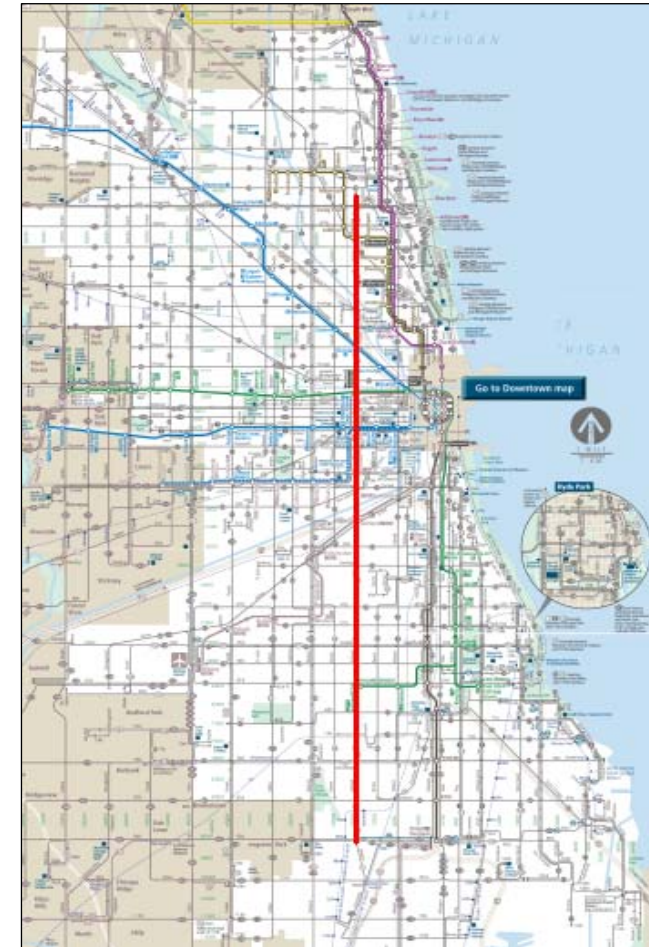
Model Components

- **Key inputs:**
 - Demand by stop
 - Bus running times
 - Limited stops
 - Frequency split
- **Key processes:**
 - Stop choice for customers closest to local only stop
 - Route choice for customers at combined stops
- **Key outputs:**
 - Travel times
 - Productivity
 - Market Share
 - Use of stops

CTA Route 9/X9 Description

X9 Implemented in Summer 2006

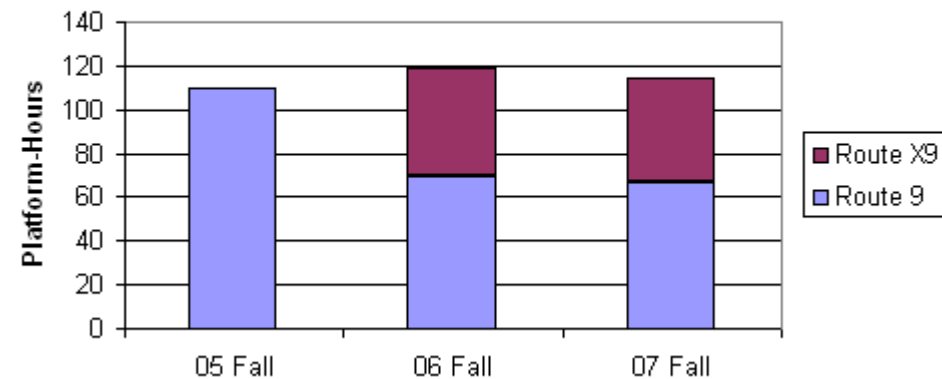
| | 9 | X9 |
|-------------------------|------------------|----|
| Length | 20 miles | |
| Average Daily Ridership | 35,000 | |
| Combined Headway | 5 min. Split 50% | |
| Number of Stops | 149 | 39 |
| Travel Time (mins): | | |
| NB | 111 | 93 |
| SB | 106 | 87 |



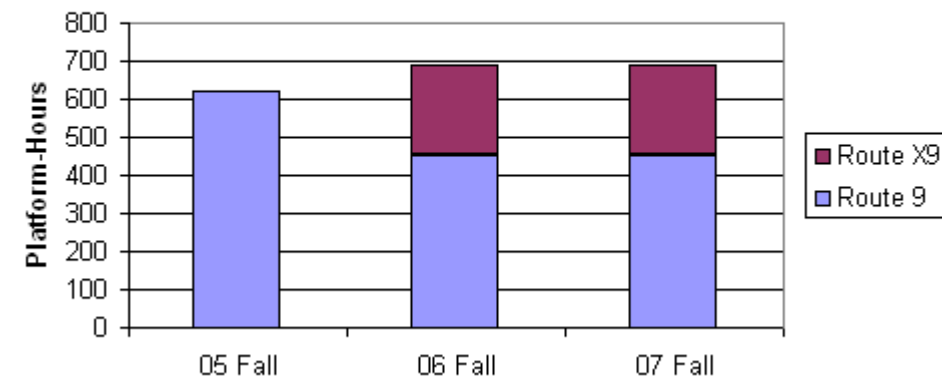
Change in Platform Hours

- **Route X9 required a modest increase in resources**

Resources (Morning peak 6:30 - 9:30)

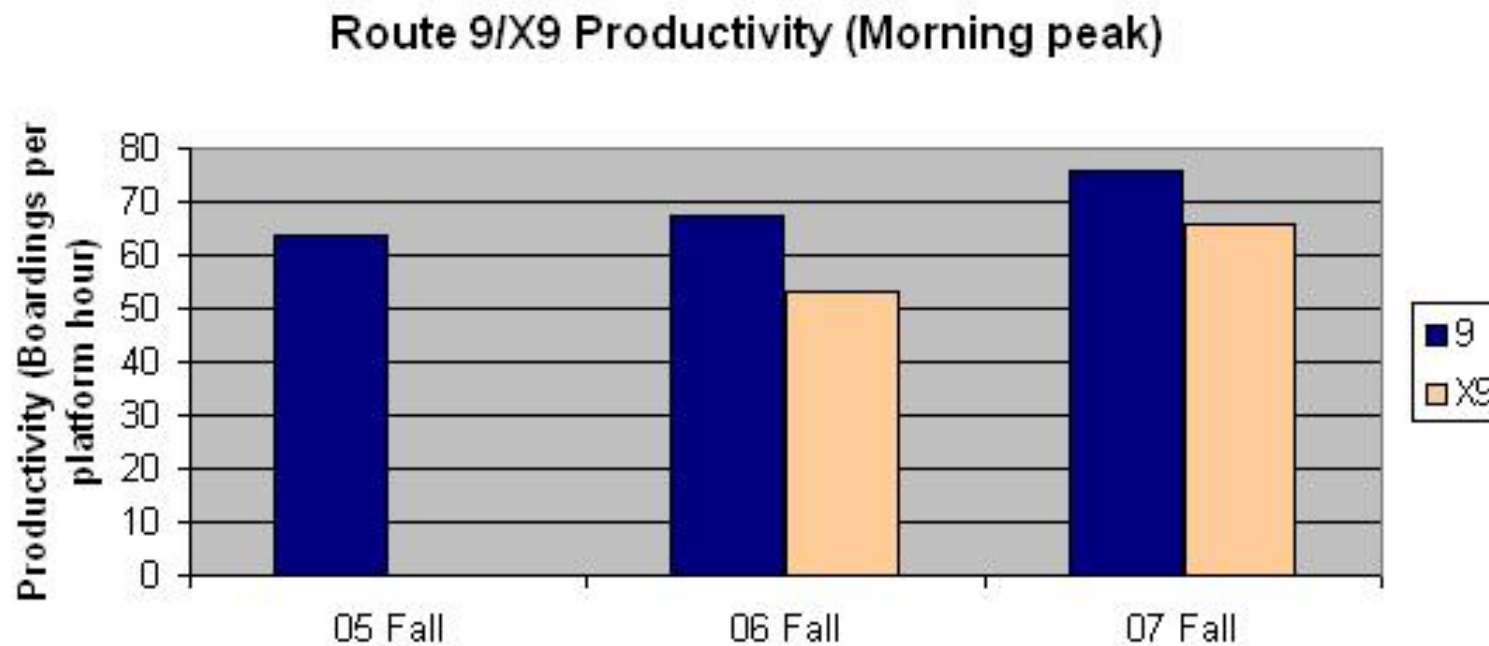


Resources (All day)



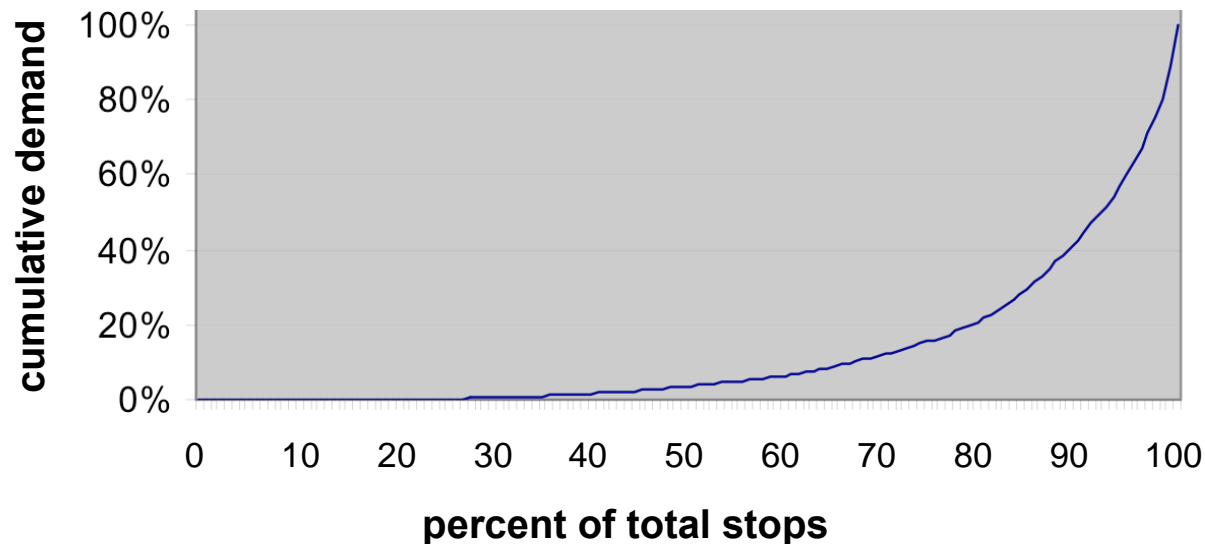
Productivity

- Overall increase in productivity



Limited Stop Design Guidelines

- **Stop spacing:**
 - CTA X services typically serve 25-30% of local stops
 - Stop spacing is 0.3-1 mile between X service stops
 - Based on cumulative demand by stop function



Limited Stop Design Guidelines

- **Running time reduction:**
 - Savings of at least 15% should be achievable
- **Limited stop frequency share:**
 - Frequency on X routes must be greater than on local service
 - Typically 60% of service should be on the X route

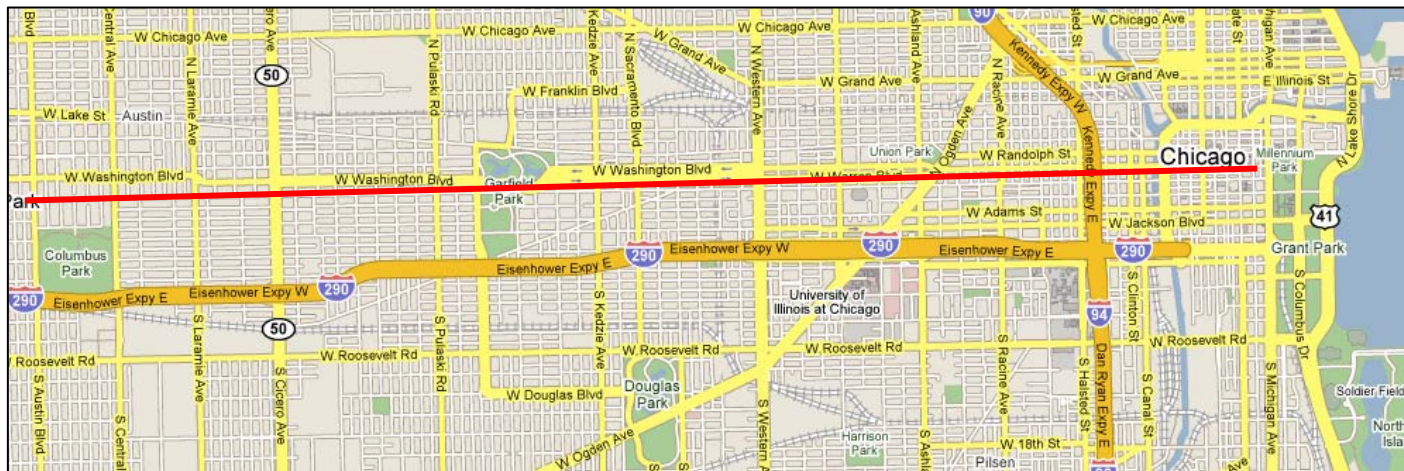
Operations Management and Control

- **Objectives:**

- Examine how real-time AVL system can support improved service reliability

- **Application:**

- CTA Route 20
- AM peak period headway: 5 minutes
- Average weekday boardings: 24,700
- 8.5 mile, 60 minute eastbound trip
- Capacity and bus bunching issues

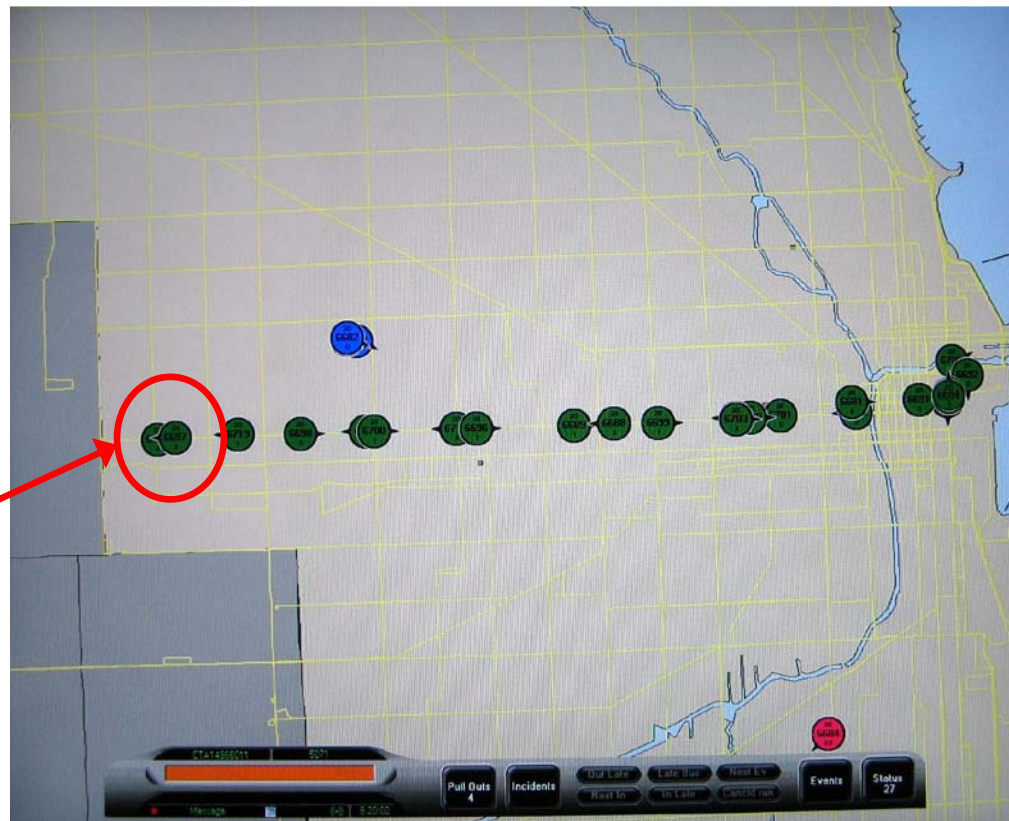


Key researchers: Chris Pangilinan (2006)

Operations Management and Control

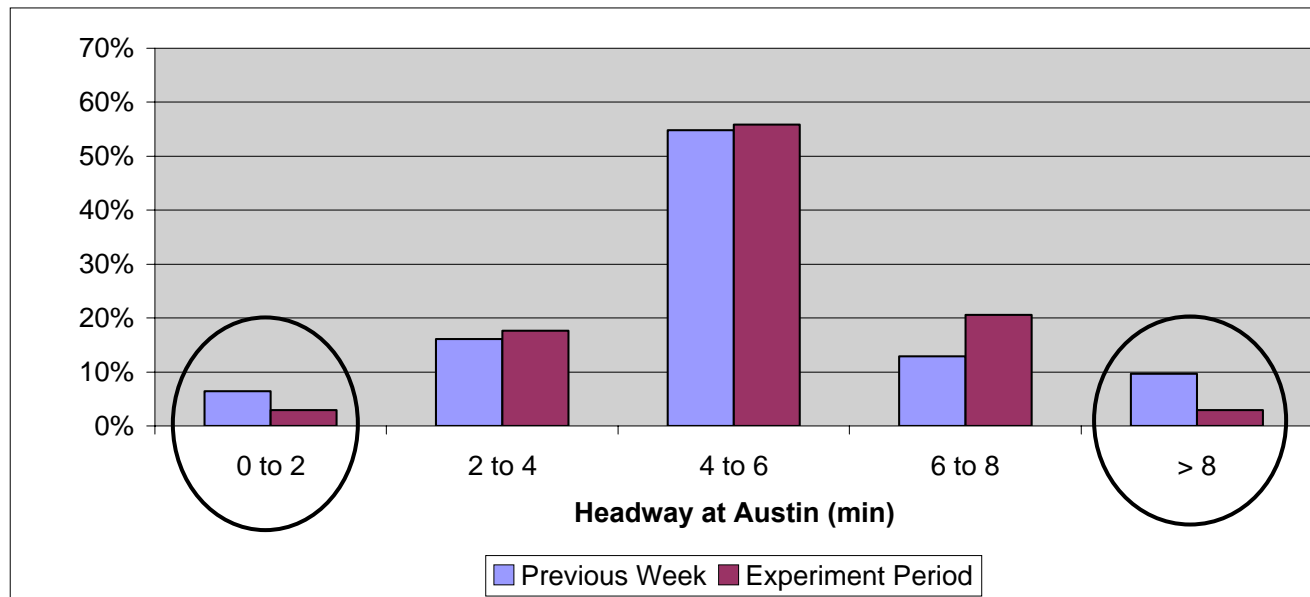
- Application:
 - CTA Bus Tracker System
 - MIT Student acted as controller for one week using Bus Tracker information

Individual Buses



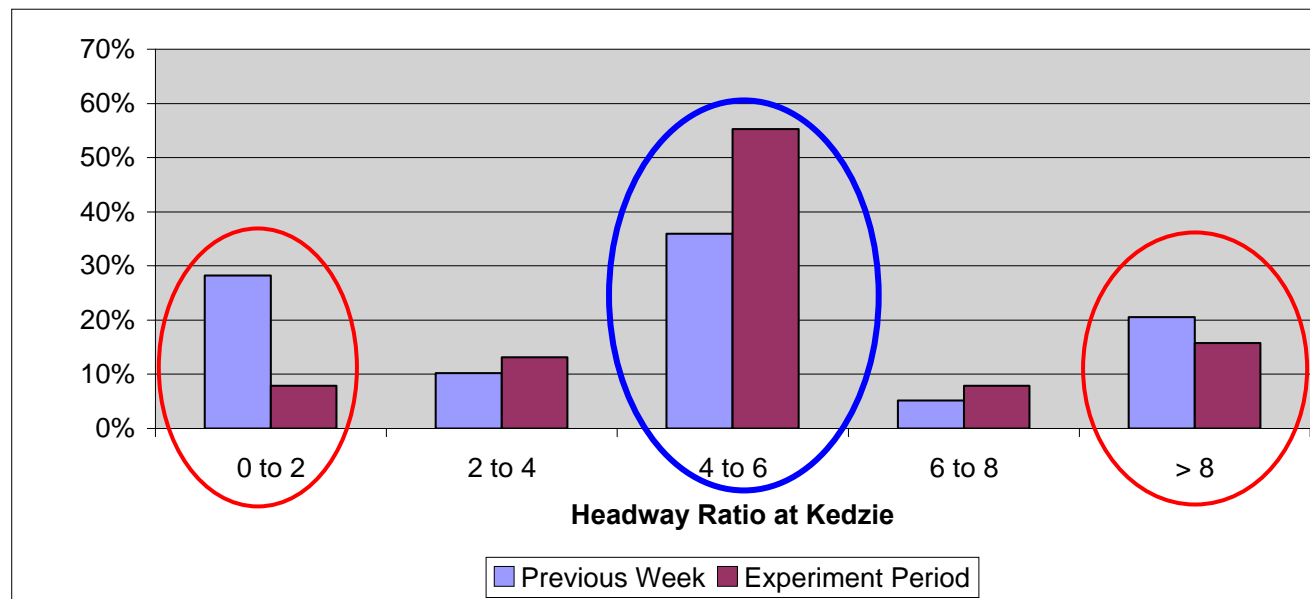
Results

- **Headways leaving Austin (Terminal)**

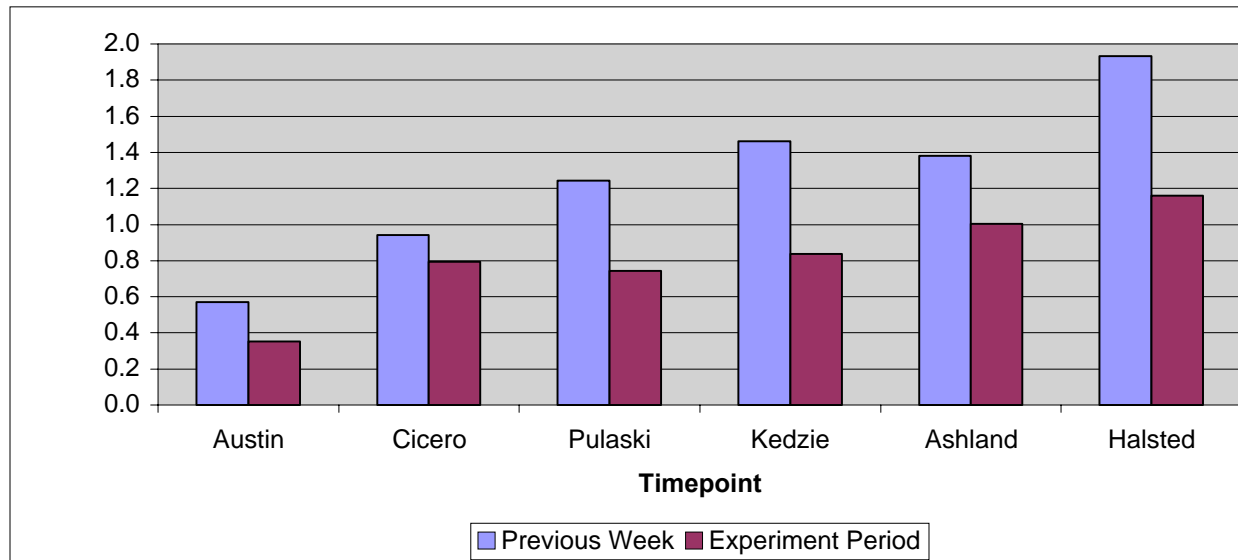


Results

- Headways 3.5 miles east of terminal



Results: Excess Wait Time



| % of Scheduled Wait Time | Austin | Cicero | Pulaski | Kedzie | Ashland | Halsted |
|--------------------------|--------|--------|---------|--------|---------|---------|
| Previous Week | 23% | 38% | 50% | 58% | 55% | 77% |
| Experiment Period | 9% | 15% | 20% | 23% | 22% | 31% |

Results in Perspective

- To achieve the same reduction in excess passenger waiting time without improving reliability:

6 more buses would have to be added to the current 24 during the AM peak

Summary of Results

- **Real-time AVL and supervision strategies contributed to:**
 - Lower rate of bus bunching
 - Lower rate of long headways
- **Downstream headways benefited from terminal control**
- **Reduction in excess wait time**
- **Dwell time, traffic, and other factors will take its toll on reliability, regardless of supervision**

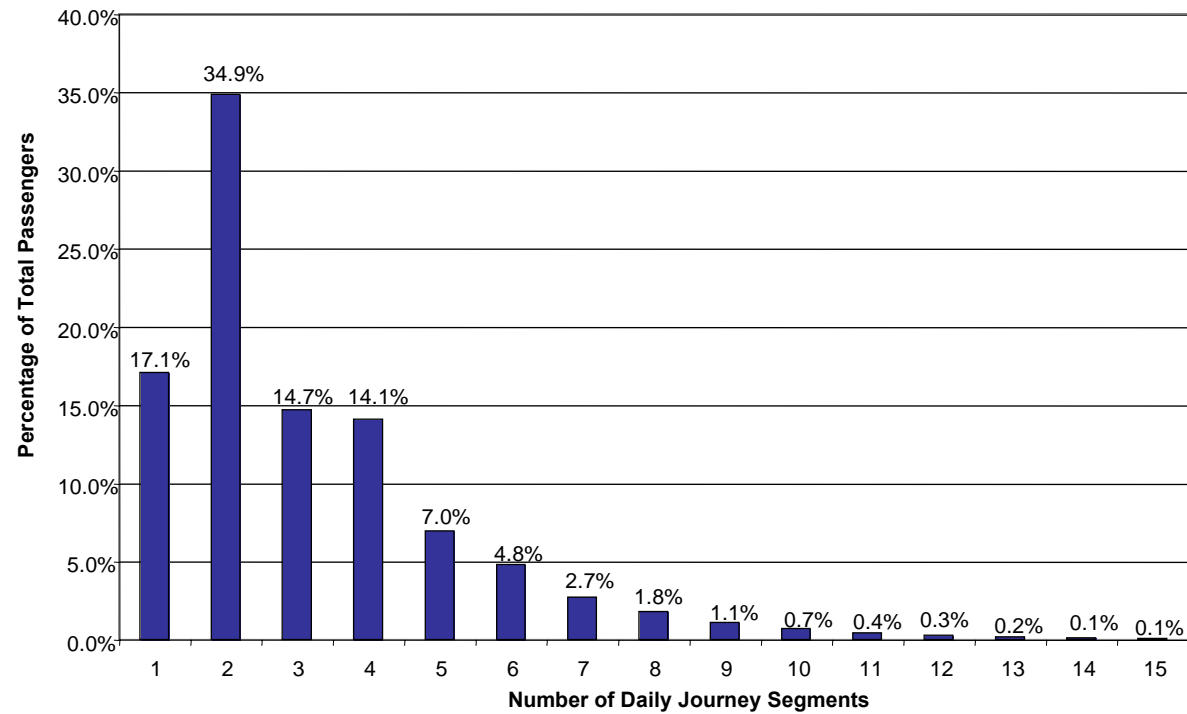
Interchange Analysis

Research Questions:

- **Can Oyster data be used to help improve the public transport network in London by focusing on bus passenger interchange behaviour?**
- **Key contribution:**
 - **Methodology for exploring passenger interchange behaviour in London using Oyster card data**

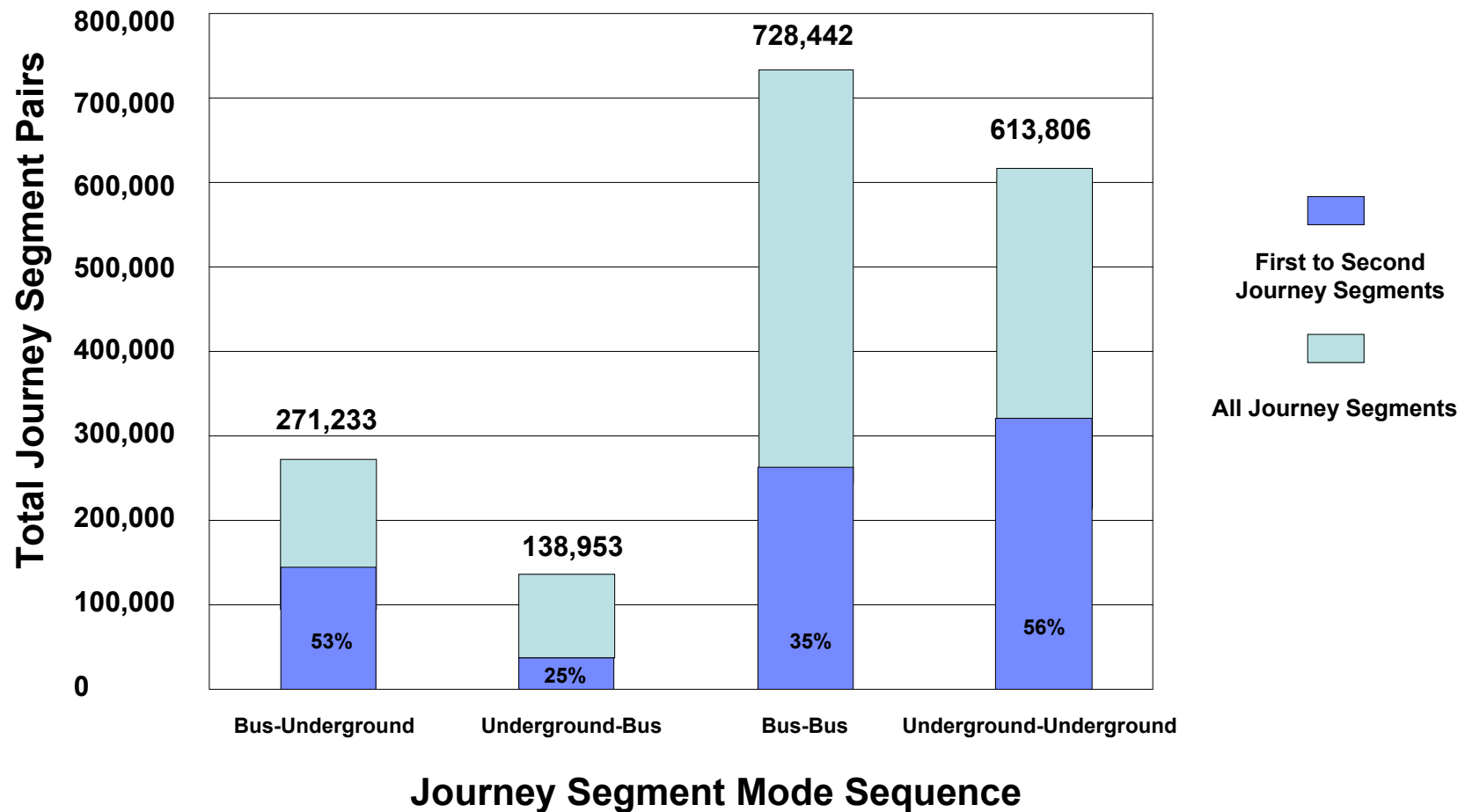
Key researchers: Catherine Seaborn (in progress)

Journey Segments Per Passenger



Source: 5% Oyster data for 2007 Period 2 (April 29 – May 26)

Consecutive Journey Segments



Source: 100% Oyster data for Wednesday, November 14, 2007

Weekday Journey Segment Patterns

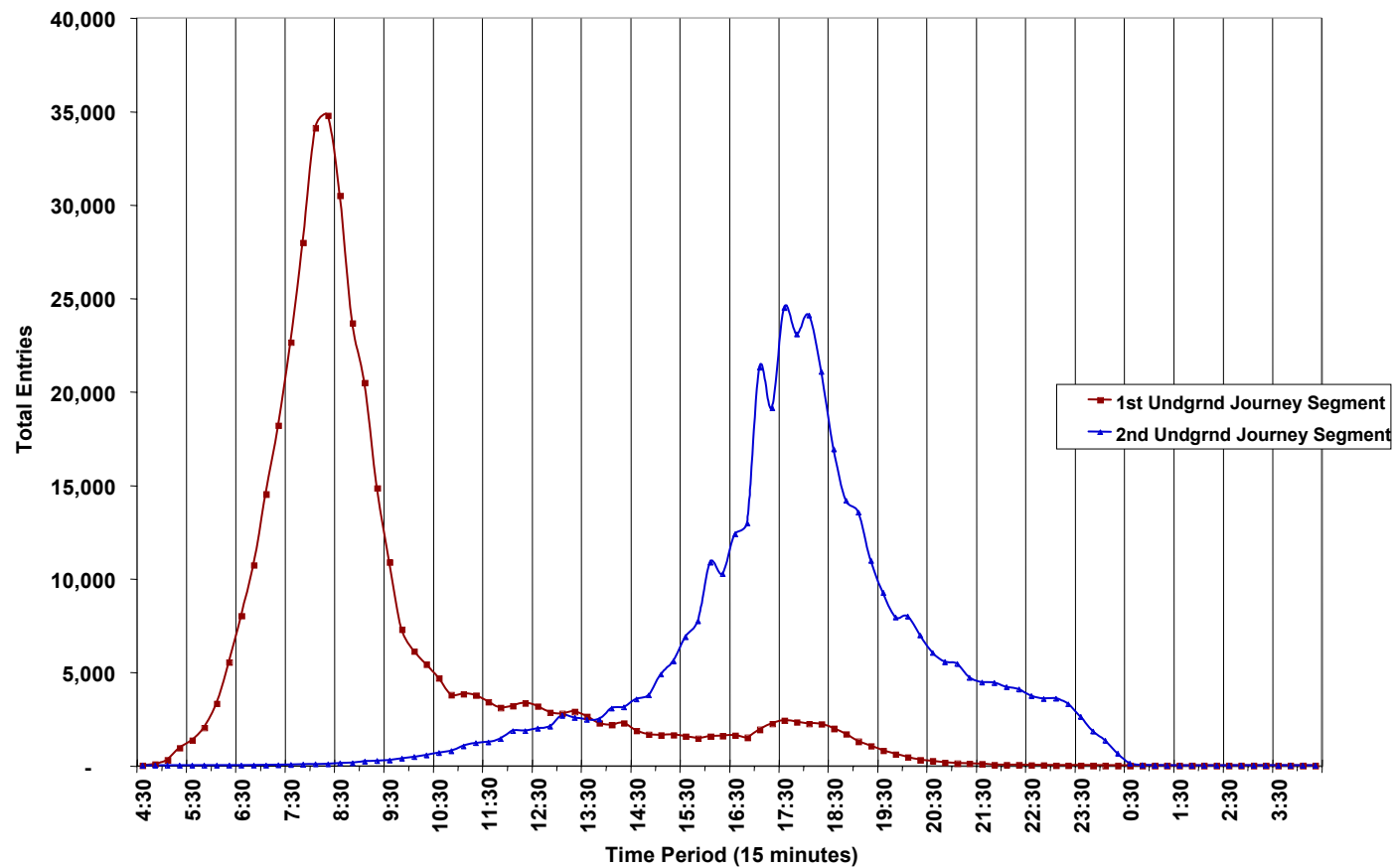
- Top 10 patterns shown

Total patterns: 15,802

| Mode 1 | Mode 2 | Mode 3 | Mode 4 | Mode 5 | Mode 6 | Passengers | Share | Cumulative Share |
|--------|--------|--------|--------|--------|--------|------------|-------|------------------|
| U | U | | | | | 416,082 | 16.3% | 16.3% |
| B | B | | | | | 401,356 | 15.7% | 32.0% |
| B | | | | | | 266,561 | 10.4% | 42.4% |
| B | B | B | | | | 150,781 | 5.9% | 48.3% |
| B | B | B | B | | | 144,275 | 5.6% | 54.0% |
| U | | | | | | 125,528 | 4.9% | 58.9% |
| B | U | U | B | | | 77,353 | 3.0% | 61.9% |
| B | B | B | B | B | | 72,943 | 2.9% | 64.8% |
| U | U | U | | | | 65,190 | 2.6% | 67.3% |
| B | B | B | B | B | B | 50,485 | 2.0% | 69.3% |

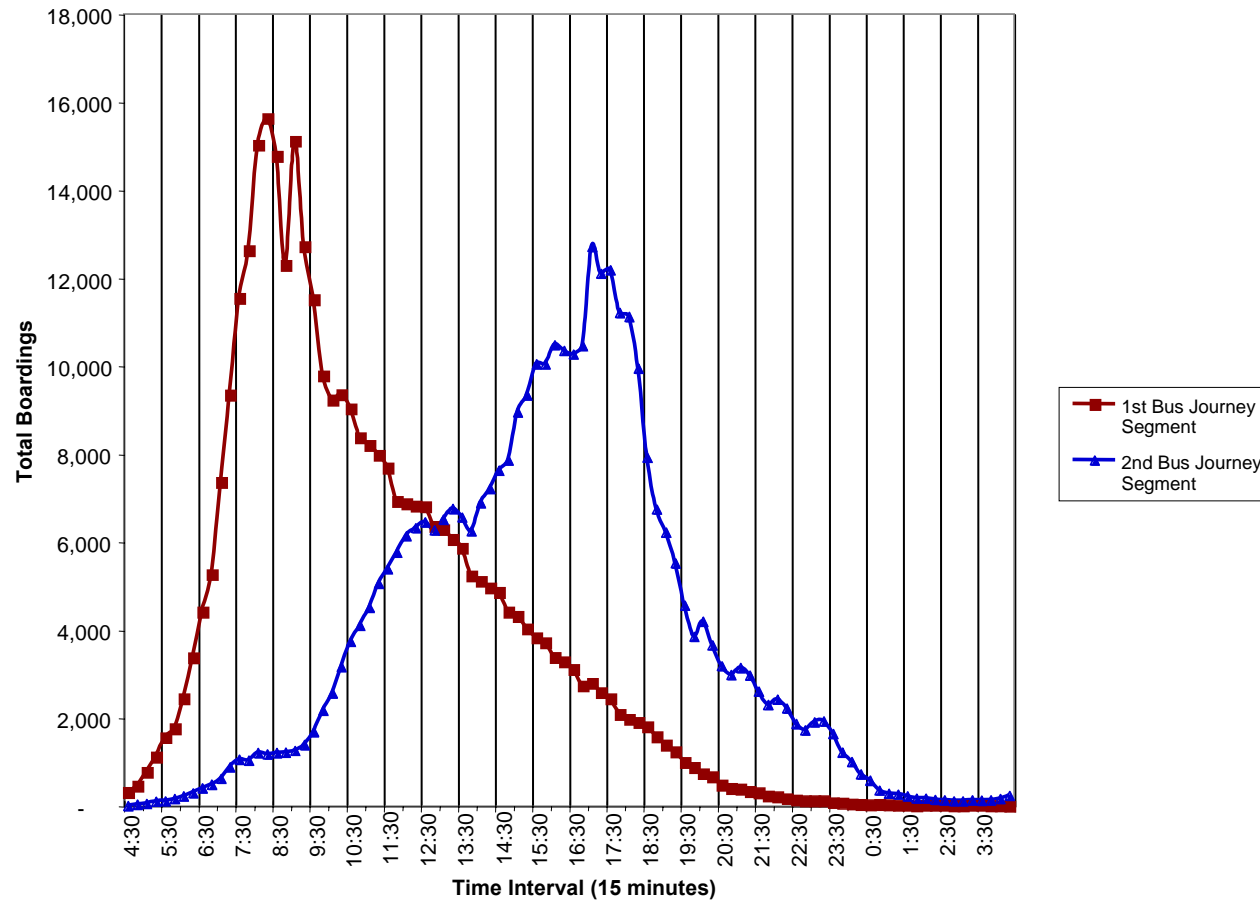
Source: 100% Oyster data for Wednesday, November 14, 2007

Underground-Underground Journey Segment Entry Times



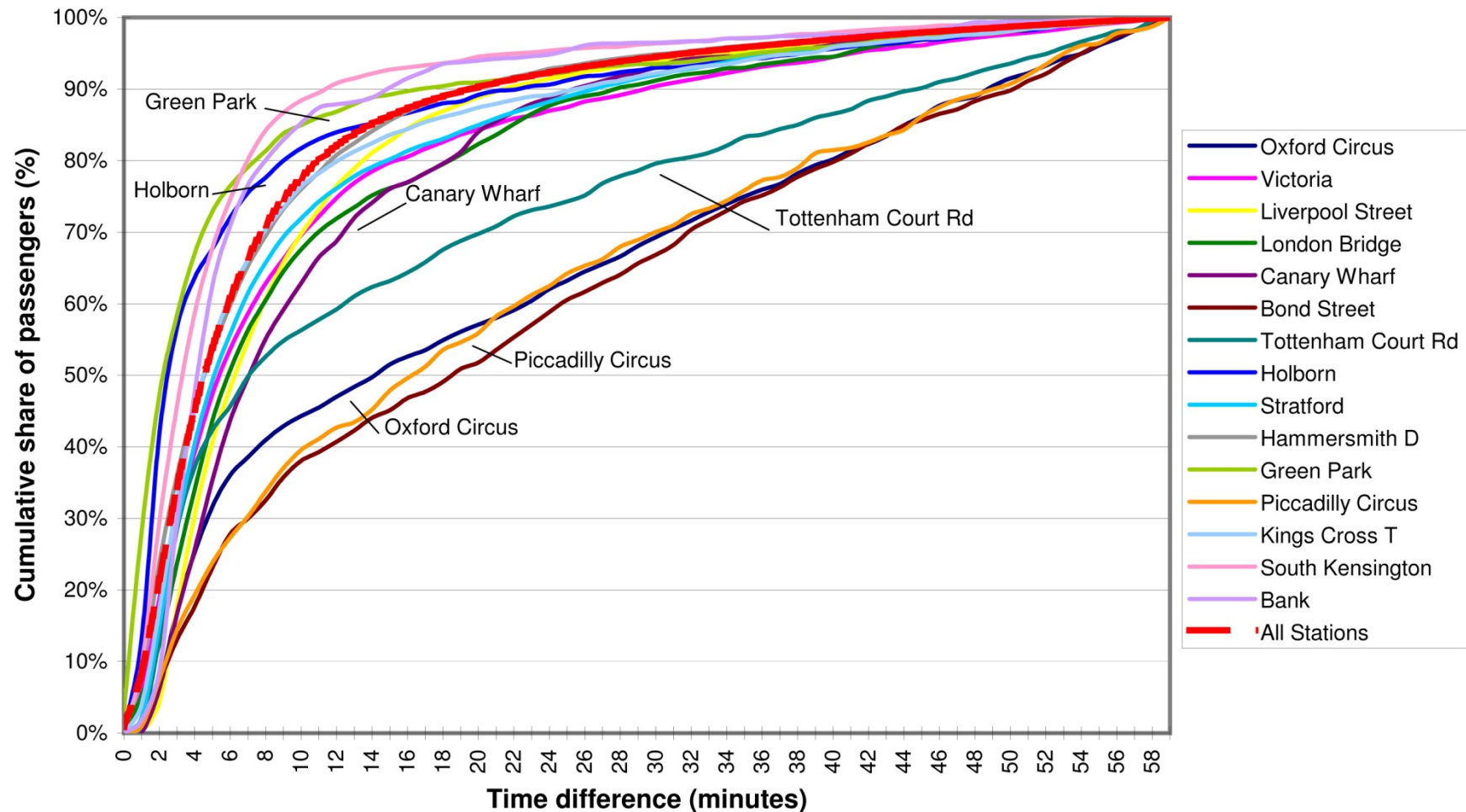
Source: 100% Oyster data for Wednesday, November 14, 2007

Bus-Bus Journey Segment Boarding Times



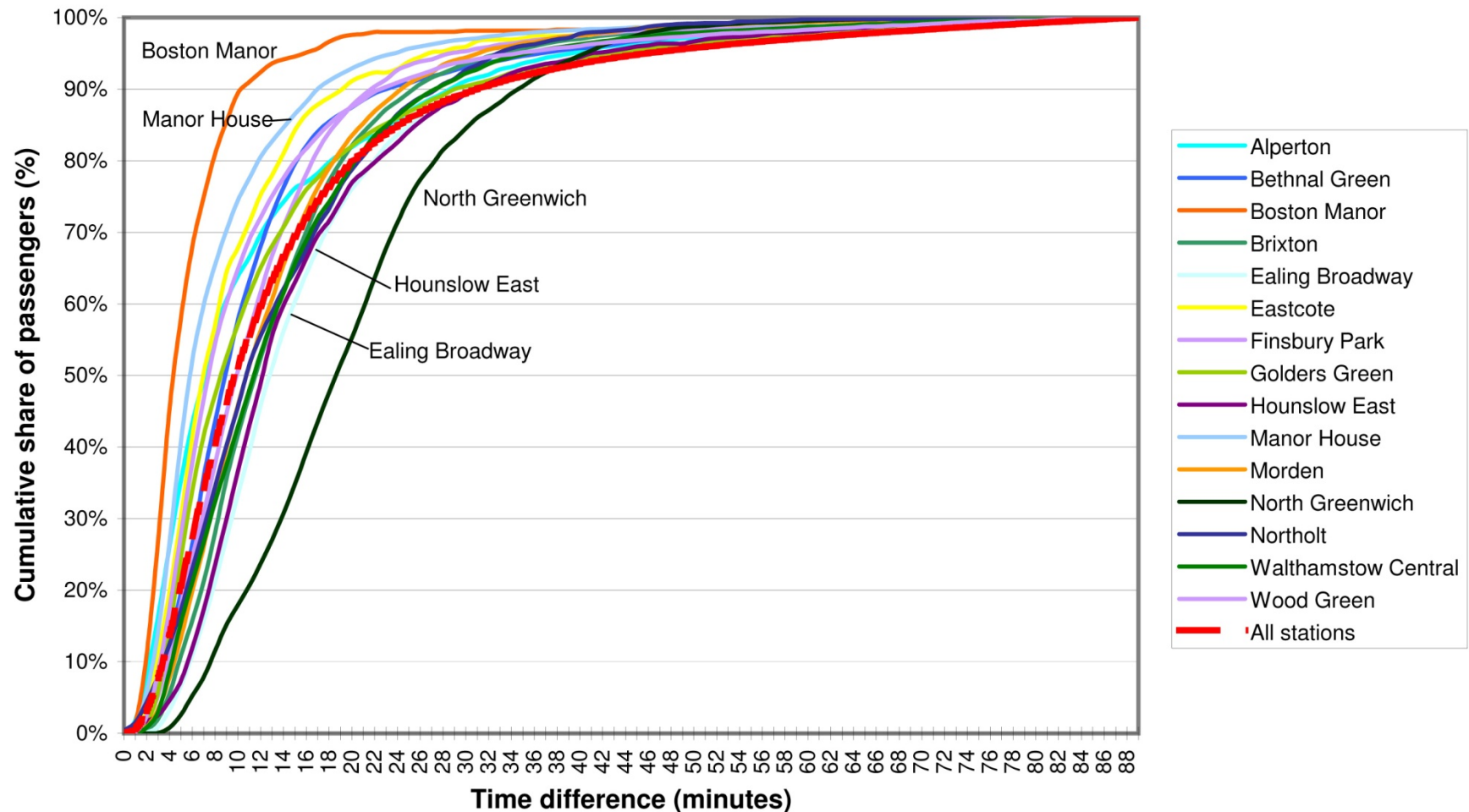
Source: 100% Oyster data for Wednesday, November 14, 2007

Underground Station Exit to Bus Boarding: Time Difference at Highest Exit Volume Stations



Data source: Transport for London. All Oyster Card journey segments on Wednesday, November 14, 2007.

Bus Boarding to Underground Station Entry: Potential Transfers are Large Share of Entries



Data source: Transport for London. All Oyster Card journey segments on Wednesday, November 14, 2007.