Measuring Transit Customers' Travel Regularity A Metric of customer behaviour using smart card data

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Introduction

Agenda

- Goal
- Methods
- Applications
- Lessons Learned



In a Nutshell

In a data driven world the core challenge we face is an abundance of data, and a lack of usable information

- Ceneral Scatter ASR: 22
 - POS Systems
 - Supply Chain Management
 - MES Systems

- Business Rules
- Strategic Decision Making
- Policy



Insights

Information Spectrum



Research Goal

To define and implement a metric that reflects a customer's tendency to make similar transit journeys over time - **Regularity**

Potential applications:

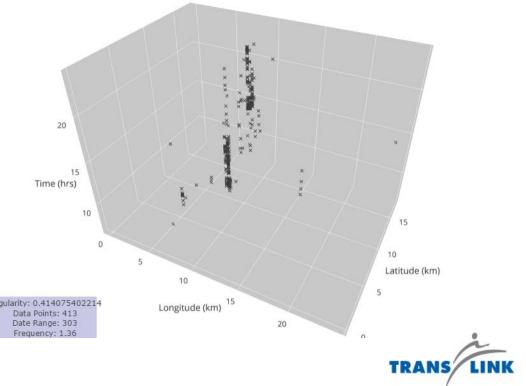
- Market Segmentation
- Pattern Detection



What is the data?

- Travel data from the smart card (Compass) system
 - Location
 - Date
 - Time
 - Mode
 - Etc.
- User specific data

Sample travel data for one person



How do we examine the data?

Research Goal

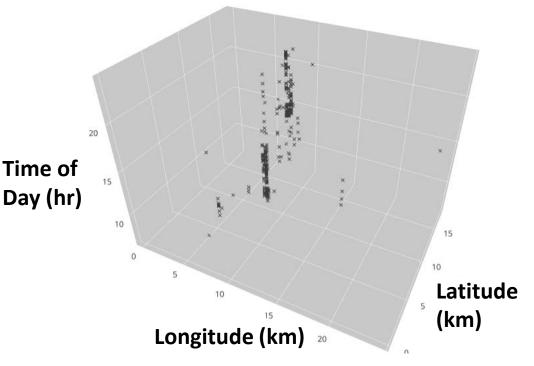
To define and implement a metric that reflects a customer's tendency to make similar transit journeys over time.



How do we examine the data?

- Reduction of data to 3 relevant dimensions
 - Origin Longitude
 - Origin Latitude
 - Departure Time
- Framing it in the Journey
 Space
- Each point represents a Journey

Sample travel data for one person



Comparing Users Journey Spaces

Transit User 1 Transit User 2 **Transit User 3** Time (hrs) Time (hrs)5 Time (hrs)6 Latitude (km) 10Longitude (km) Longitude (km) Léngitude (km) Latitude (km)

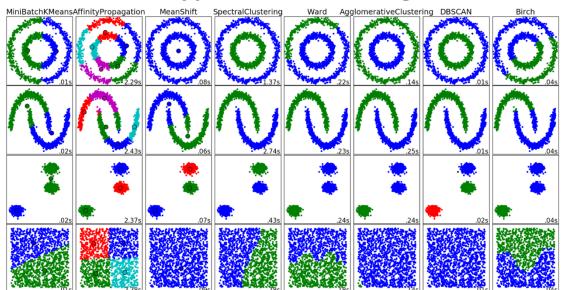
Increasing Regularity

Methods

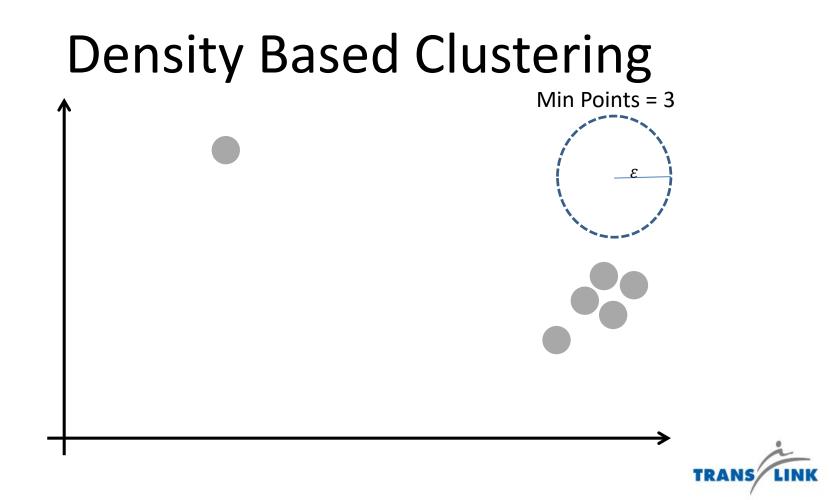
Algorithmic Clustering

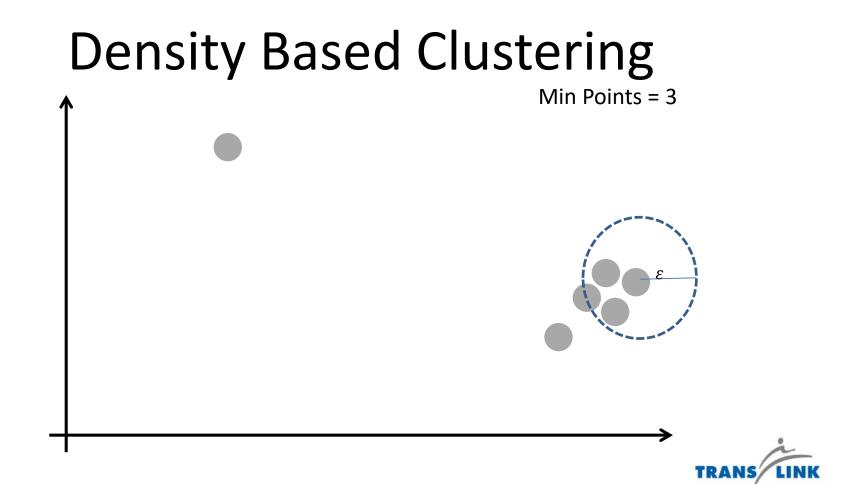
Definition: **Clustering algorithms** are a subset of machine learning tools that group a set of data points into sets (clusters) of similar objects based on similarity of specific parameters

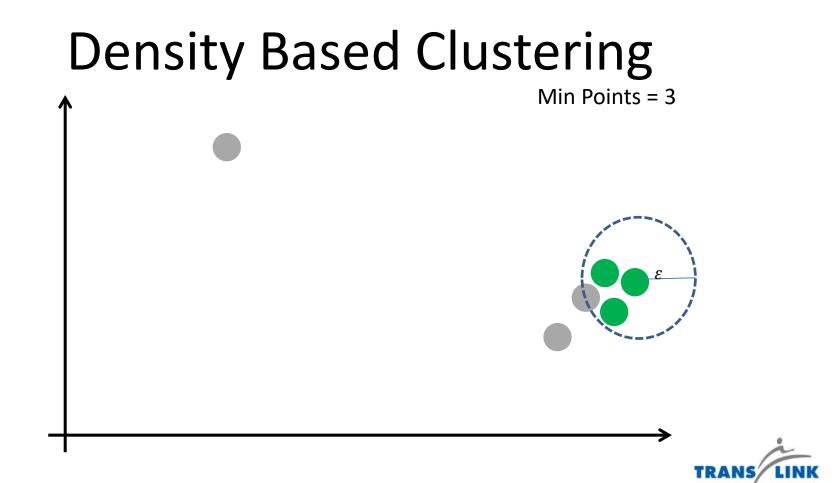
Types of Clustering

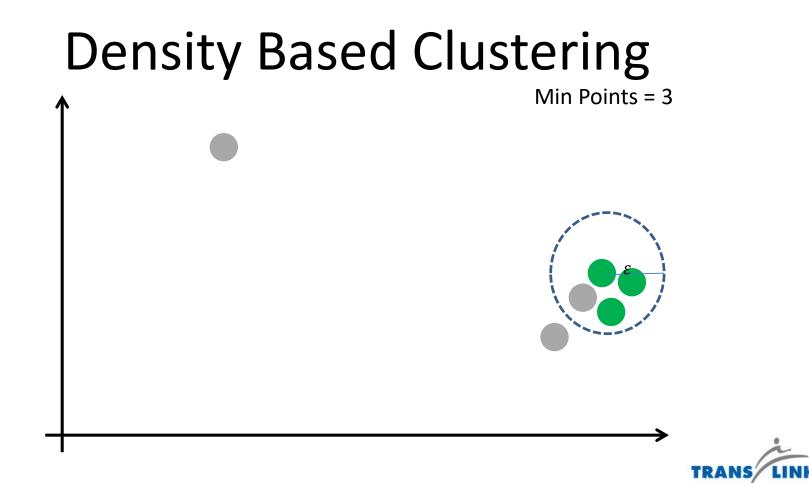


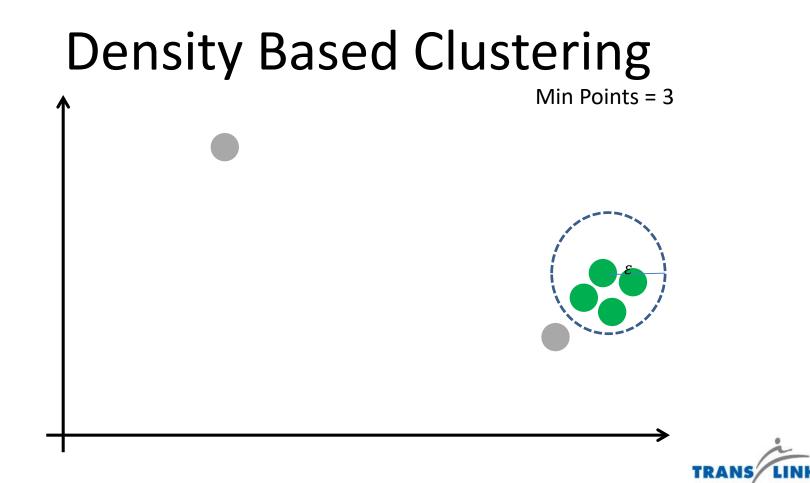


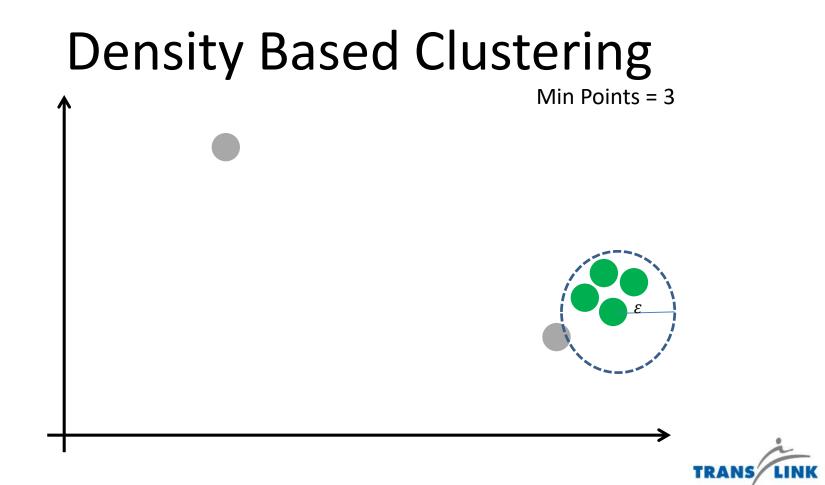


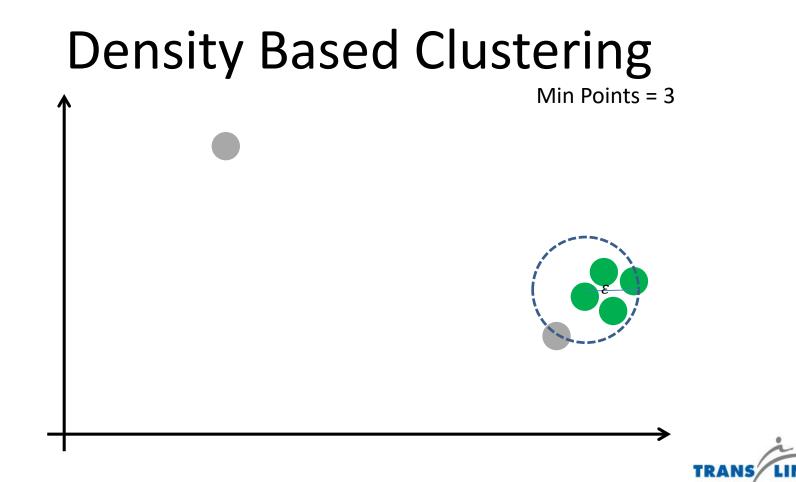


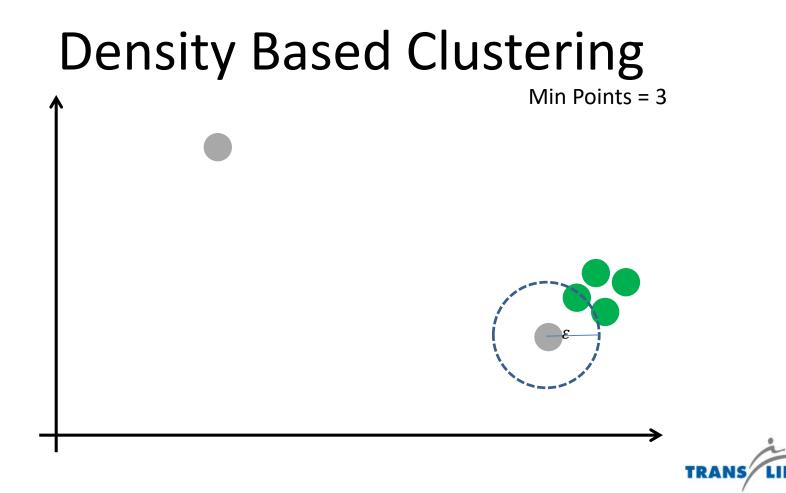




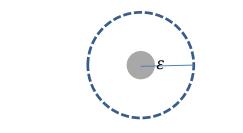








Density Based Clustering

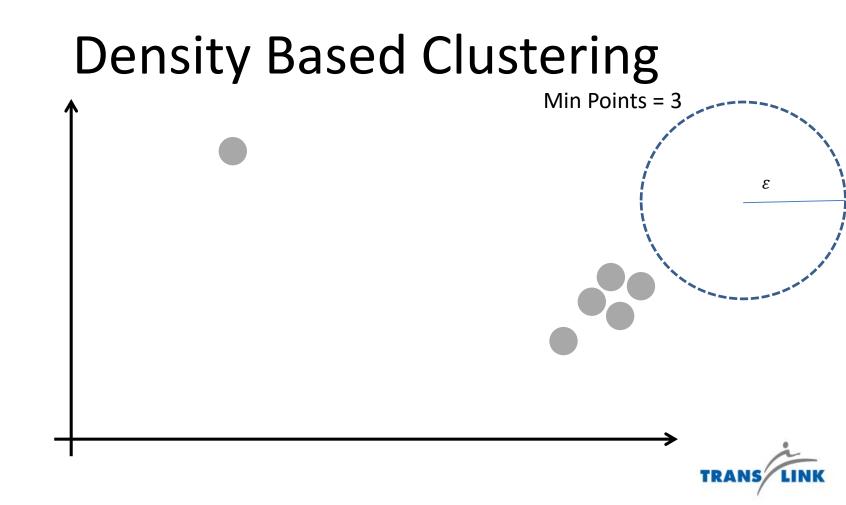


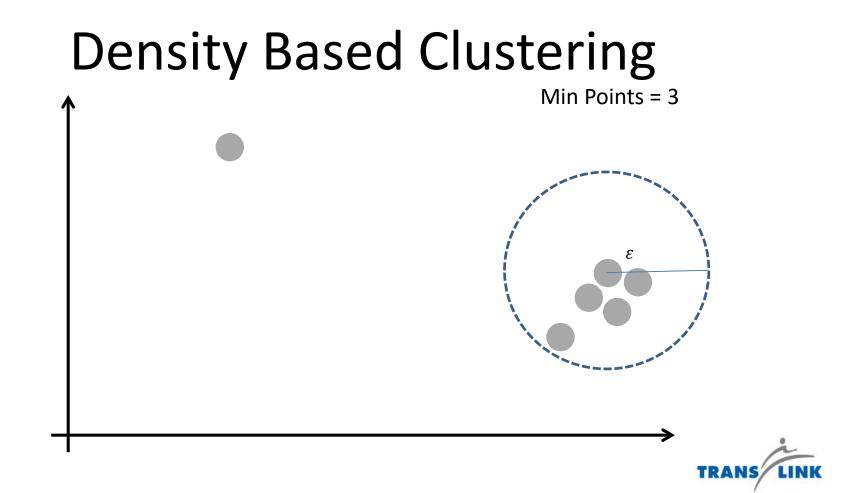
4 Clustered points + 2 noise points

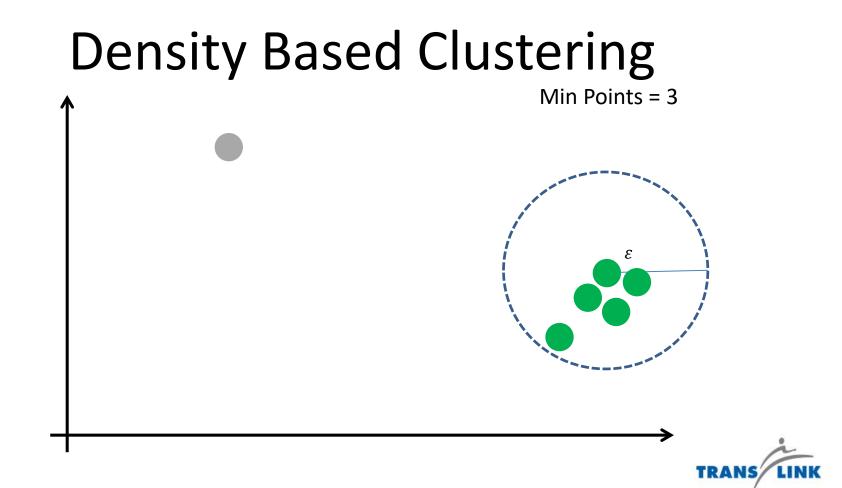


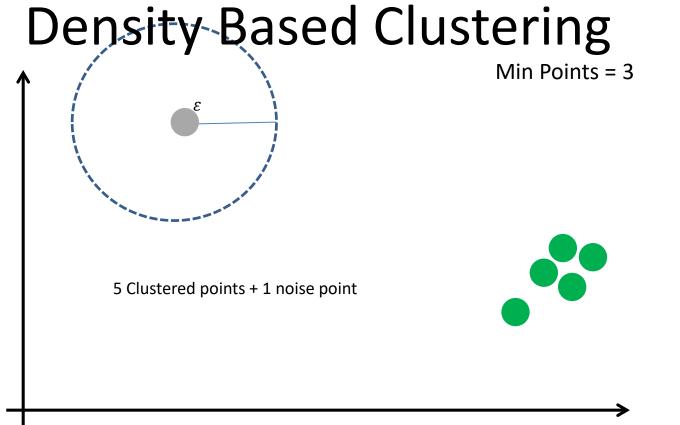
Min Points = 3



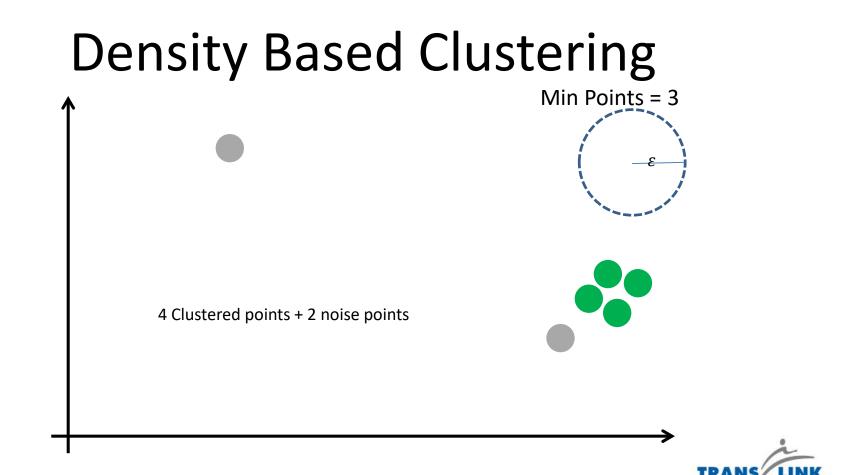


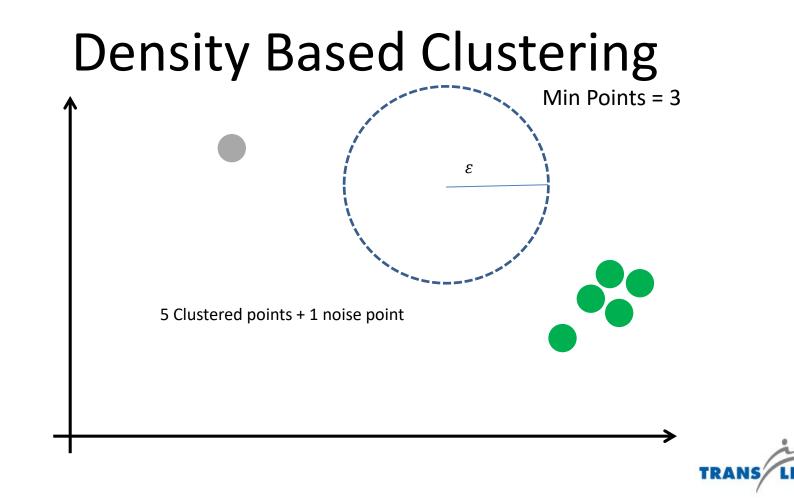




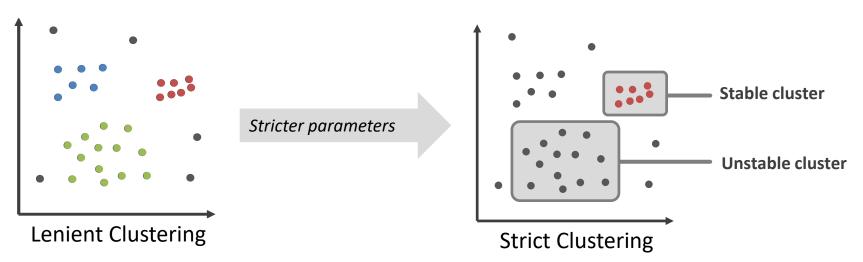








Cluster Stability



Regularity: customer's tendency to make *similar journeys* over time

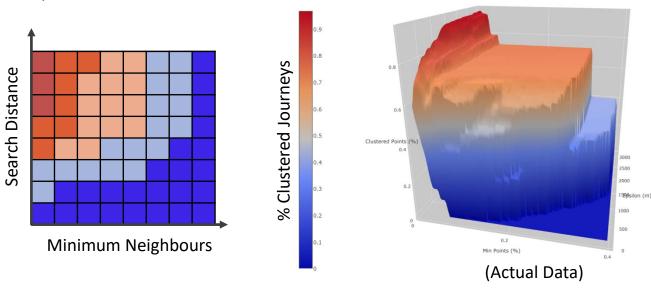
If a large proportion of a customer's journeys fall into clusters, AND these clusters survive strict parameters, the customer can be considered "regular".

Methods – Calculating Regularity

Selected clustering algorithm (DBSCAN) has 2 parameters:

- Search distance
- Minimum neighbours

To measure regularity for a card, we run DBSCAN with various parameter combinations...

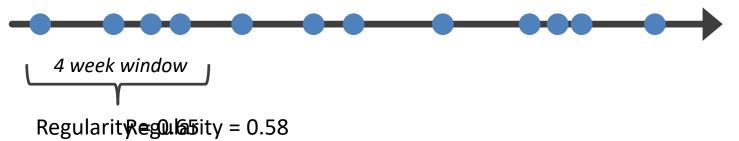


Regularity = average % of journeys that fall within clusters.



Regularity Over Time

How does regularity change over time?



	Period					
Card	1	2	3	4	5	6
A	0.65	0.58	0.55	0.48	0.58	0.6
В	0.70	0.75	0.72	0.65	0.60	0.53
С	0.20	0.25	0.27	0.29	0.30	0.30

Findings & Applications

Trends in Regularity

Average Regularity by Period

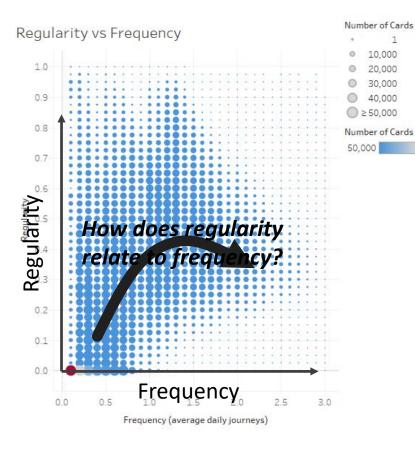


Regularity declined through first 8 months of 2017, then recovered in the Fall.



Regularity vs. Frequency

3.128.373



- Wide variation in regularity for all frequencies.
 - Regularity tends to increase with frequency, up to ~1.3 journeys per day, then declines.



Findings – Unexpected Transit Market

Regularity

Where do Stored Value and Monthly Pass users sit within this spectrum?

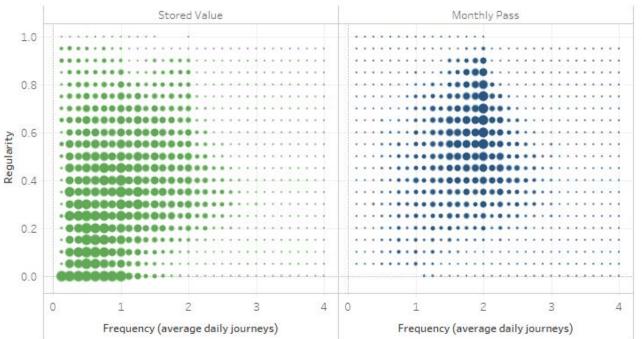
Conventional wisdom is that Stored Value users are "irregular" users.



Frequency

Regularity vs Frequency

Regularity vs Frequency by Product



Number of Cards

0.0	1	0	30,000
0	10,000	0	40,000
0	20,000	0	≥50,000

- For a given frequency, Monthly Pass users tend to be more regular than Stored Value users.
- There is a large contingent of Stored Value users who travel at lower frequencies, but with high regularity.

Future Applications

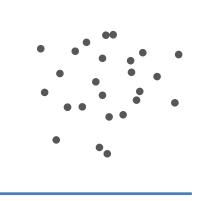
- Market Segmentation
- Pattern detection / behaviour trends



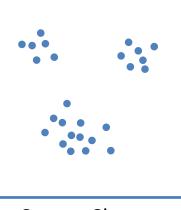
Thank you!

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No clusters (all noise)



Strong Clusters