



Innovation, Science and  
Economic Development Canada  
Communications Research Centre Canada

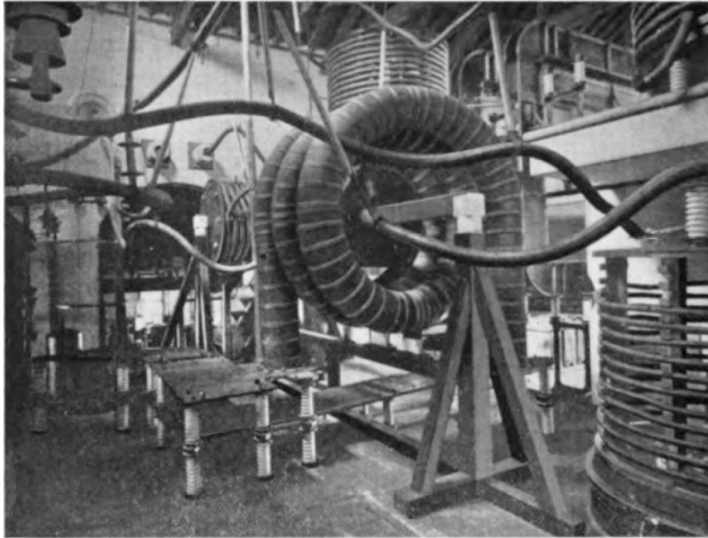
Innovation, Sciences et  
Développement économique Canada  
Centre de recherches sur les communications Canada

# Dynamic Spectrum Management

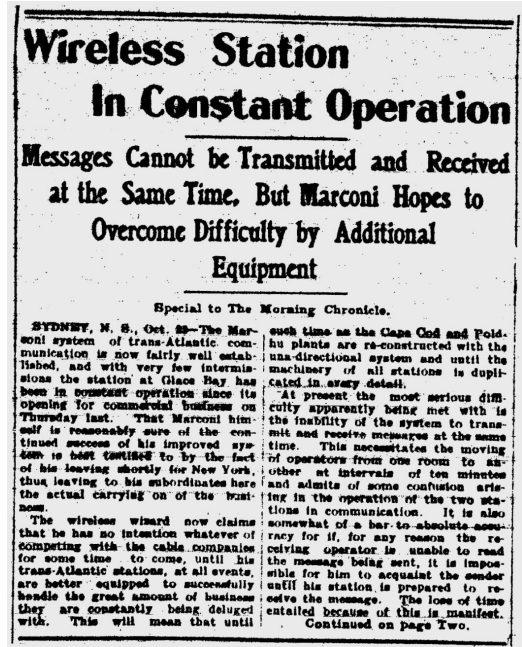
Amir Ghasemi  
November 14, 2018

Source: Communications Research Centre Canada

# 100+ Years of Interference (Avoidance)

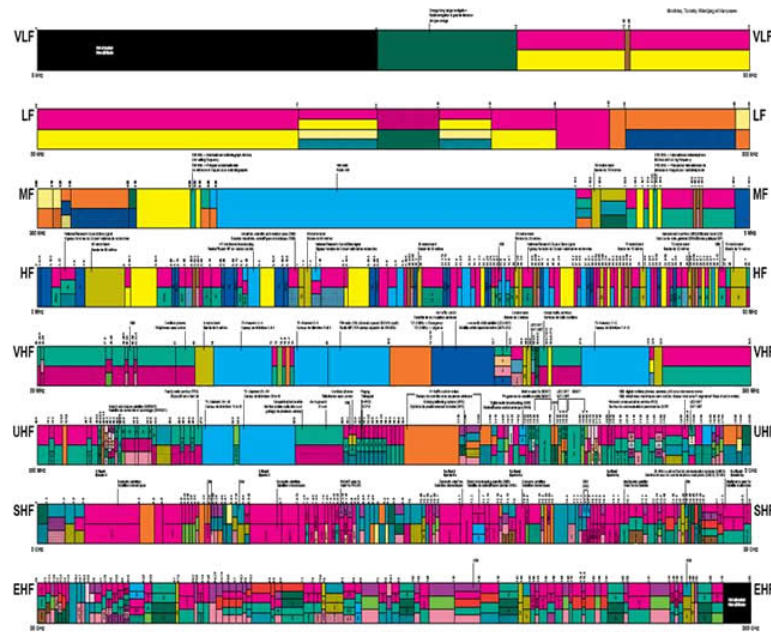


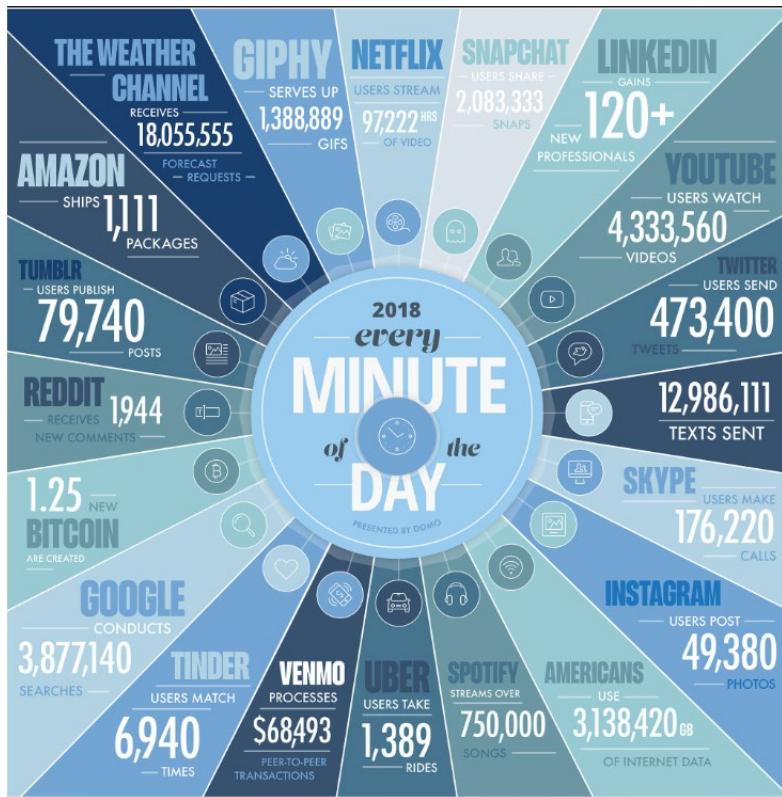
Marconi's trans-Atlantic transmitter



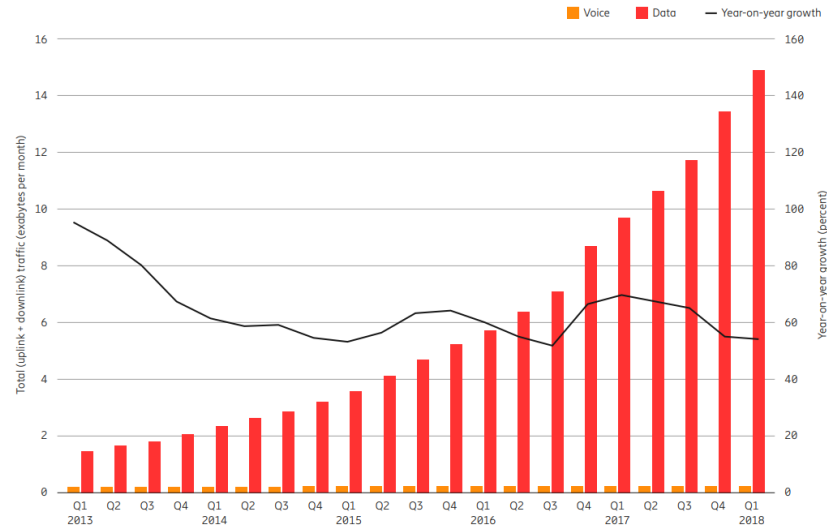
# Overview

- Spectrum management today:
  - Hard guarantees via resource fragmentation and conservative assumptions
  - Manual, labour-intensive processes in some bands
  - Inefficient under time and location varying loads

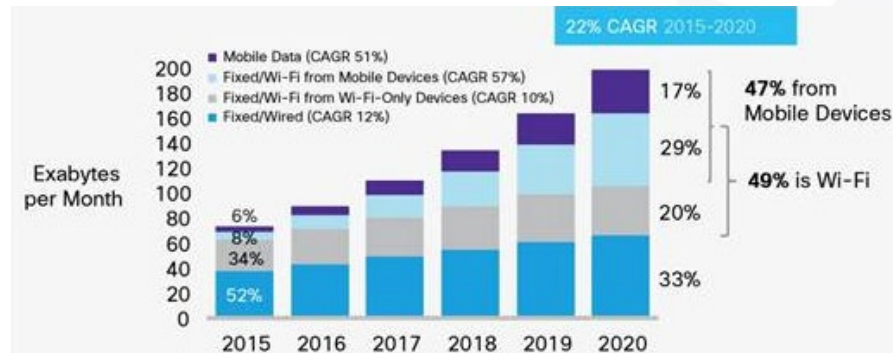




source: Domo Inc.



source: Ericsson (Q1 2018)



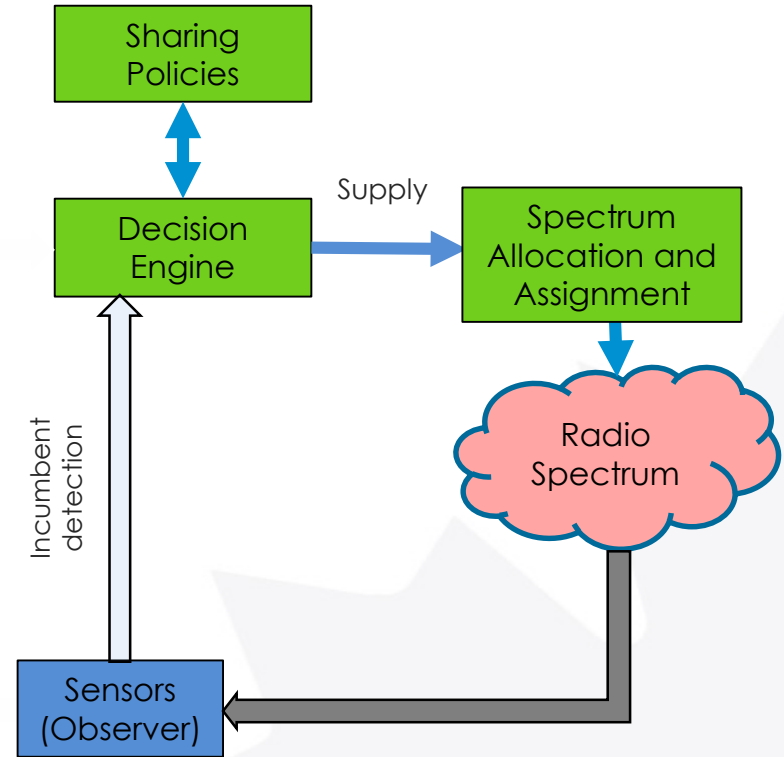
source: Cisco (2017)

# Spectrum Outlook

- More mobile spectrum pressure
- mmWave shows some promise but bound by laws of physics
- Better use of spectrum below 6 GHz still needed
  - New technologies, repacking, auctions

# Towards More Dynamic Management

- Sharing with incumbent protection (TVWS, CBRS)
- Spectrum “supply” determined by incumbent activity

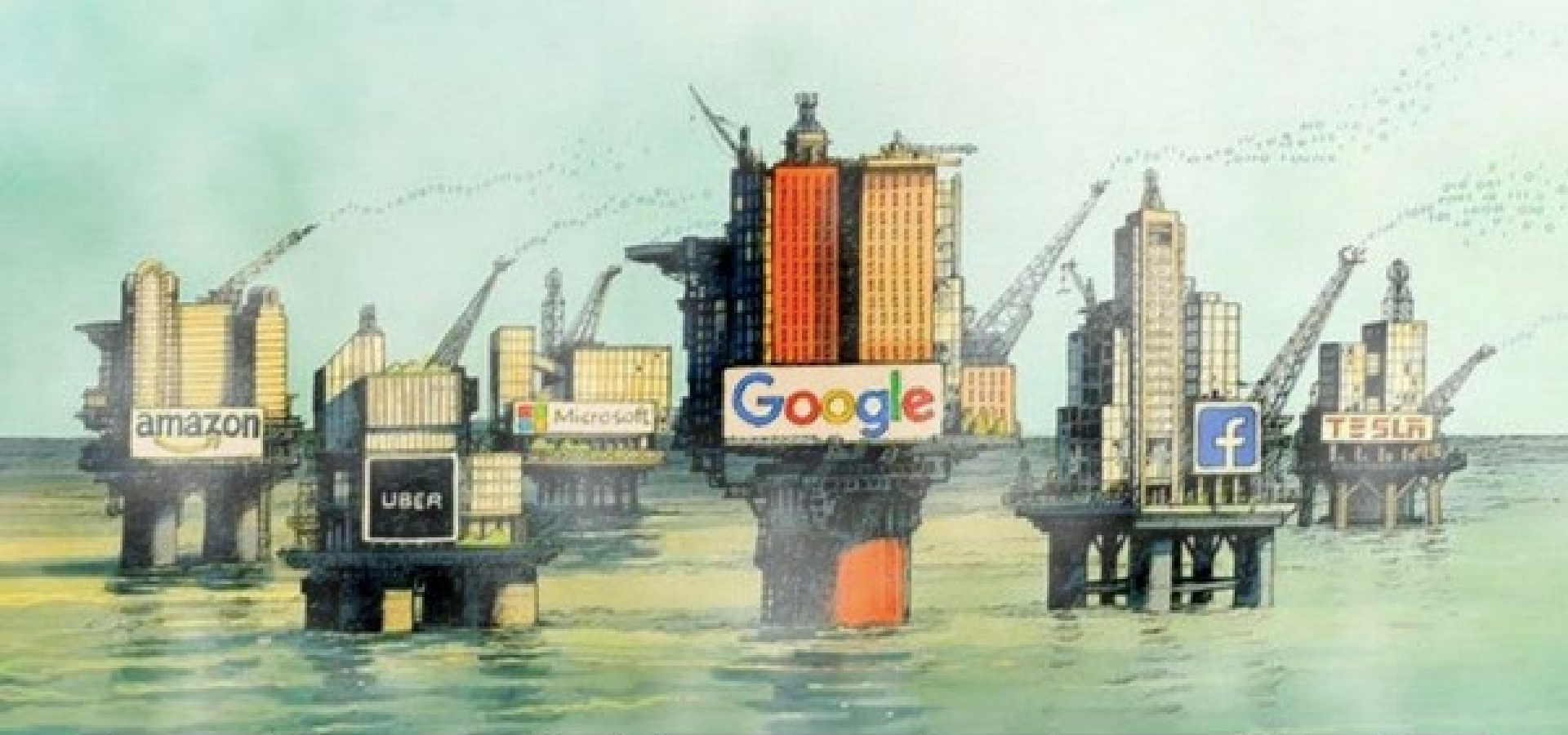


# Dynamic Spectrum Management

- How can we manage increasing **demand** by sharing bands among different technologies?
- **How**, **when**, and **where** spectrum is being used?
- What policies would promote **efficiency** of systems sharing the spectrum?

**Challenged by lack of data to model usage, predict demand, and measure success**





The world's most valuable resource is no longer oil, but data.

The Economist - May 2017

David Parkins



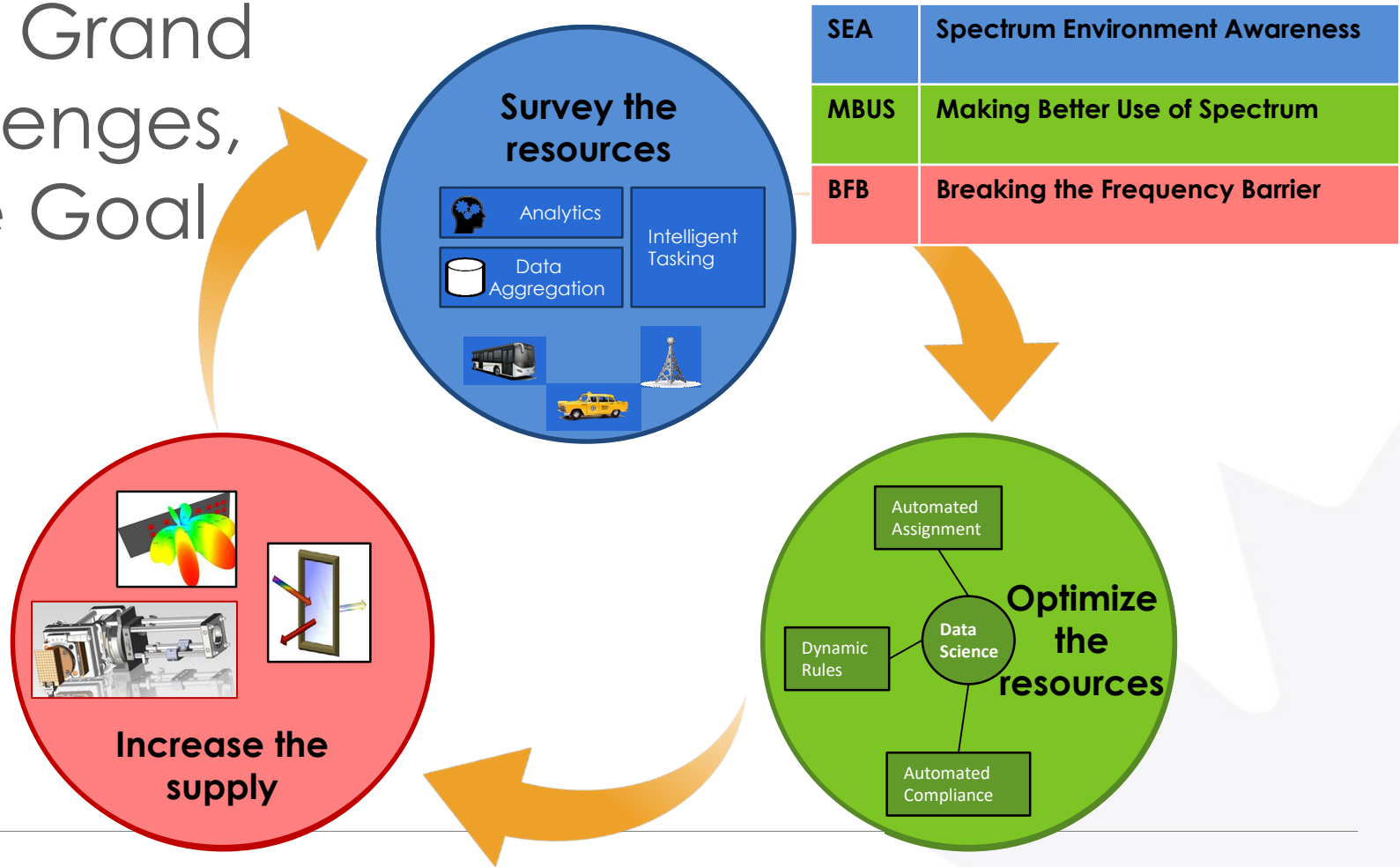
# Communications Research Centre Canada (CRC)

Federal government's primary R&D lab for advanced telecommunications

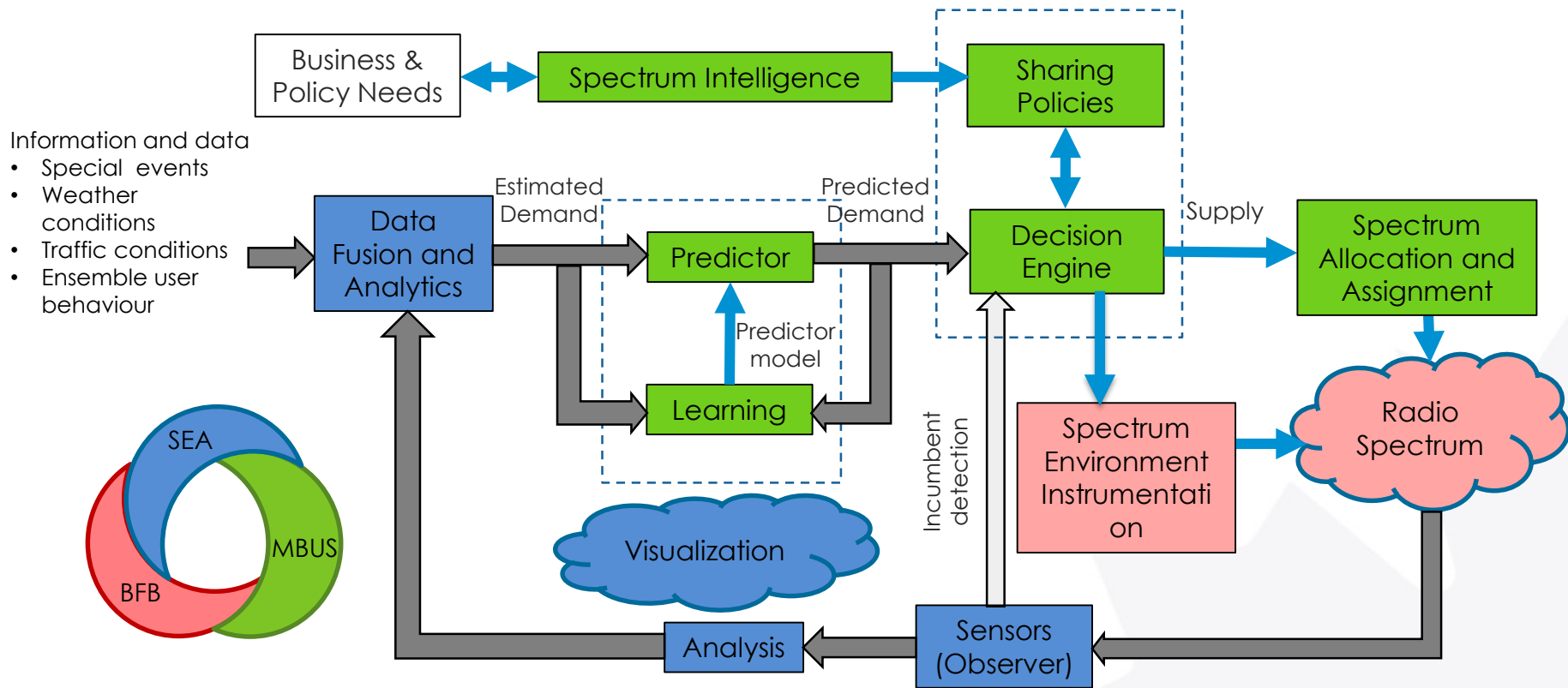
- A client-driven applied research centre focused on “what is possible and what works”



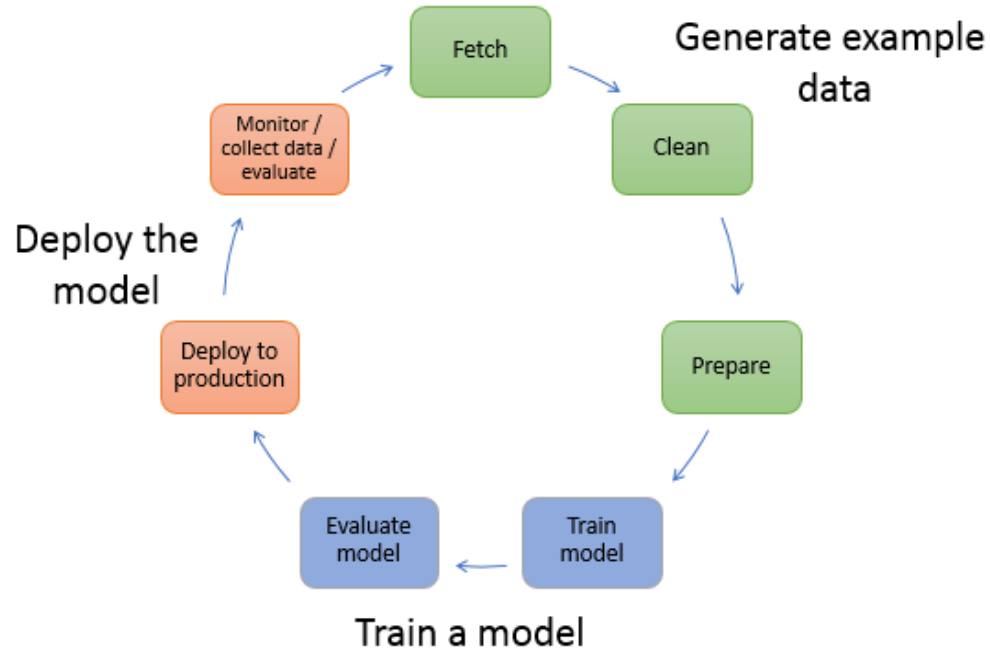
# Three Grand Challenges, One Goal



# Grand Challenge Framework



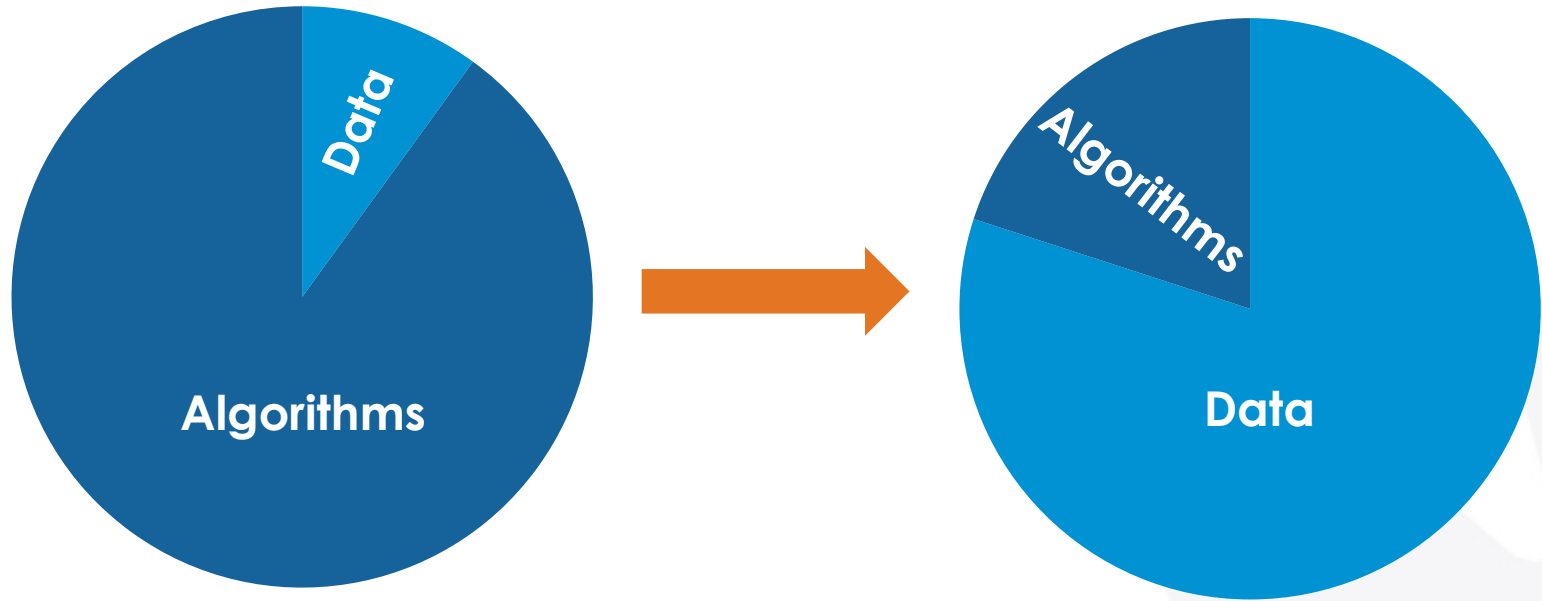
# Generic Machine Learning (ML) Cycle



# ML/AI in Wireless Communications

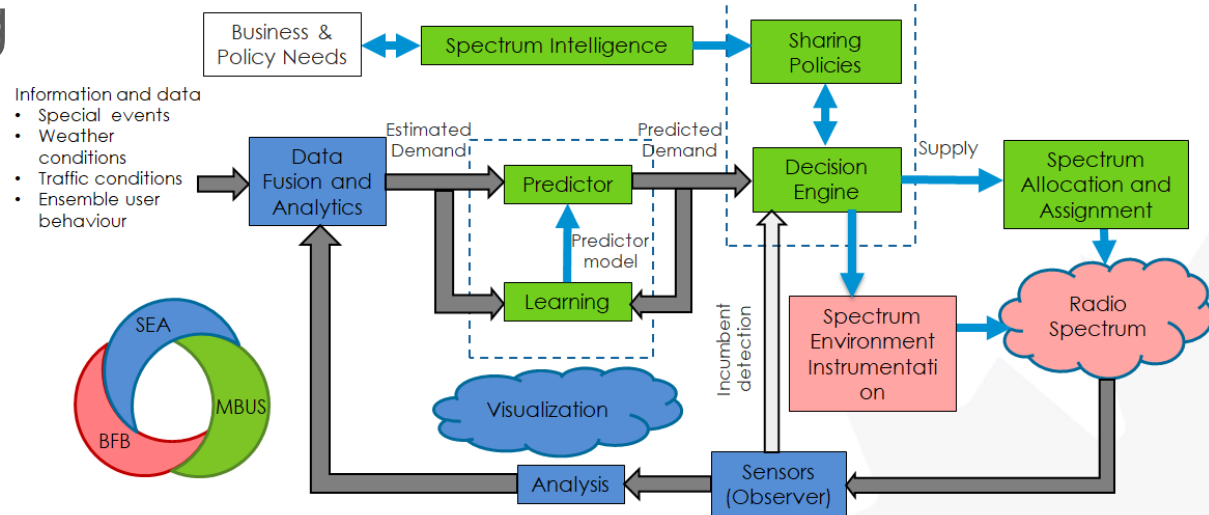
- Academia: signal processing, source/channel (de)coding, modulation recognition, etc.
- Manufacturers: network optimization, scheduling, rate control, anomaly detection
- Operators: mostly relying on manufacturers and 3<sup>rd</sup> parties so far
- Spectrum regulators: ?

# Evolution toward Automating Evidence-Based Decisions



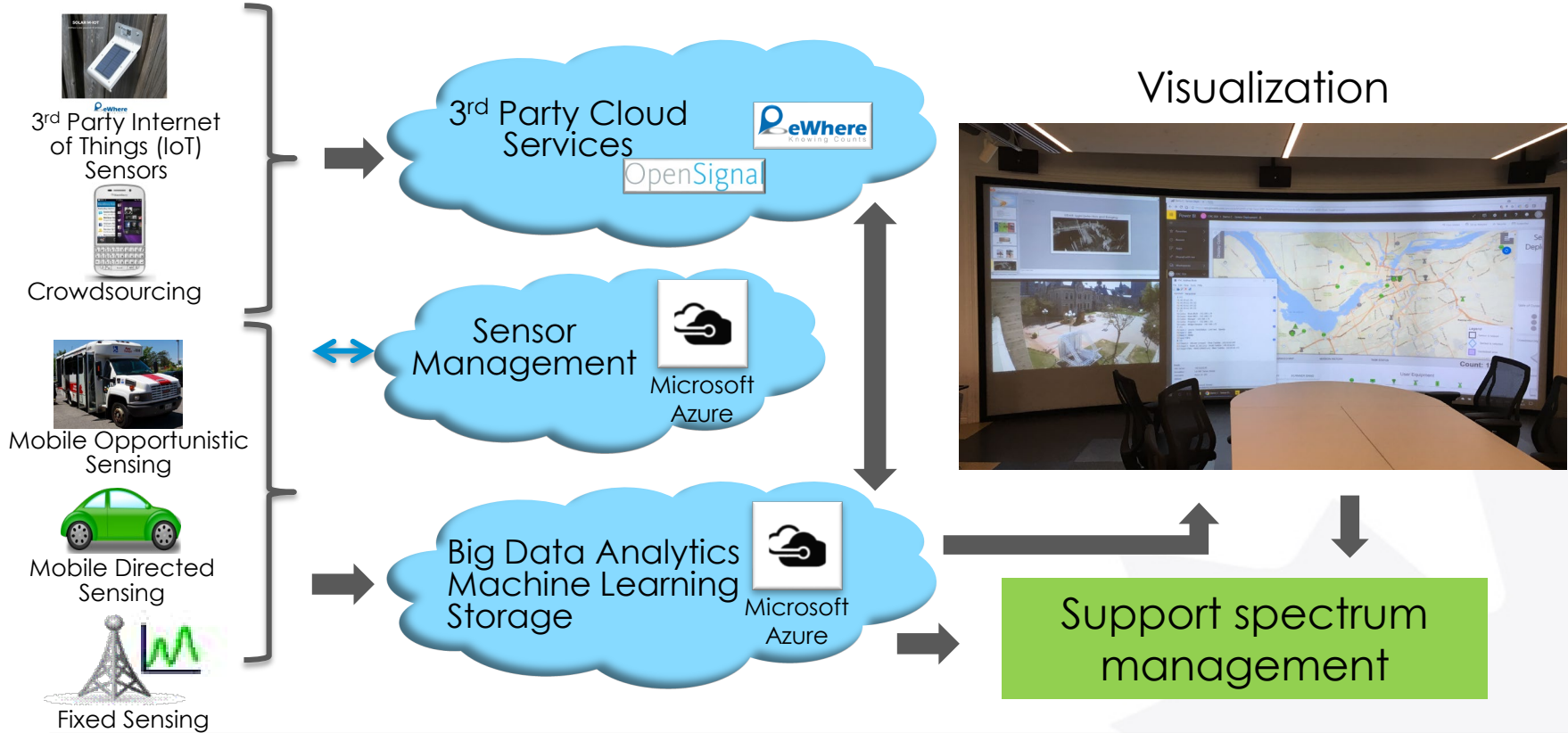
# Spectrum Usage Data Acquisition

- Direct sensing
- Operator data
- Crowdsourcing





# Cloud – Sensors and Data Collection

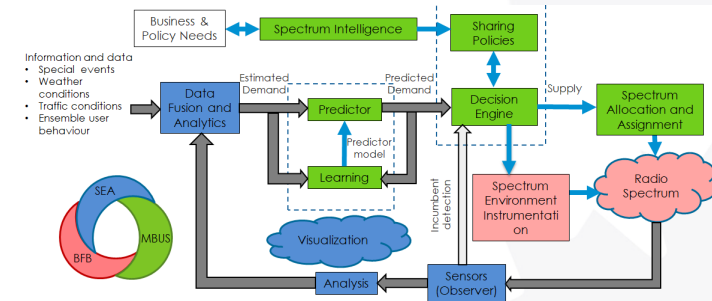


# Cloud – Sensors and Data Collection

- 110 sensors located everywhere in Canada
- ~30 advanced sensors, each uploading 30GB of raw data to Azure cloud daily
- Automated nightly cluster computing processes format data for downstream analytics tasks

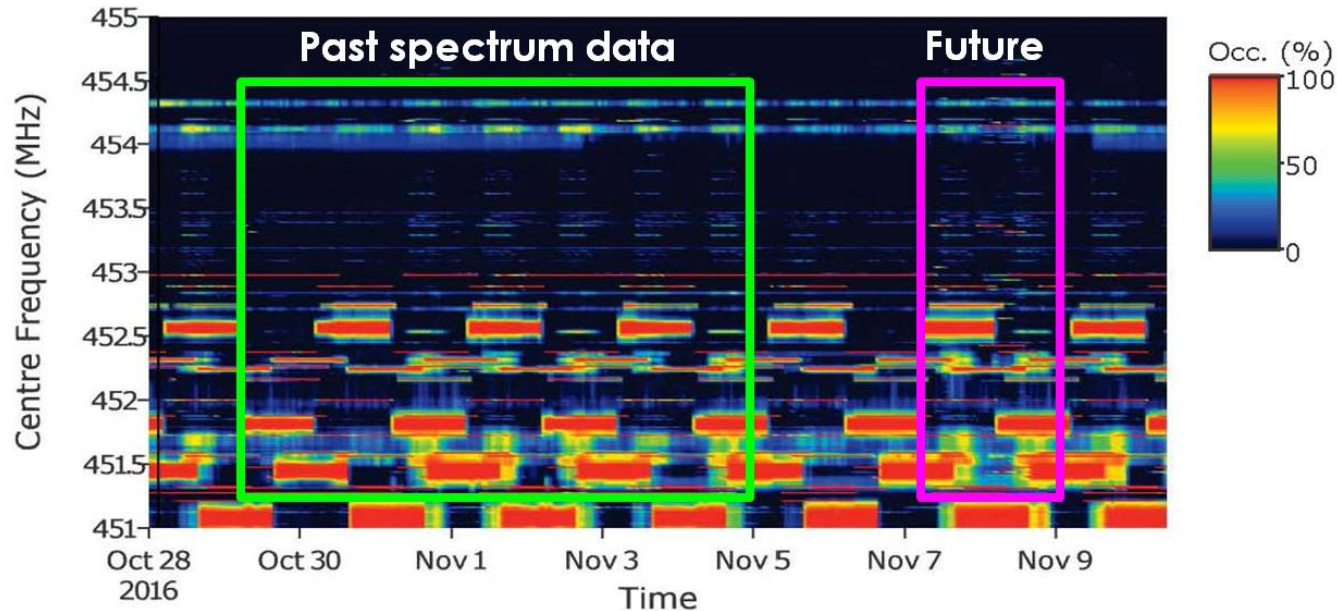
# From Data to Knowledge

- Stages of data analytics
  - Descriptive (e.g., traffic modelling)
  - Predictive (spectrum occupancy prediction)
  - Prescriptive (dynamic spectrum assignment)
- Complex multi-dimensional problems
- AI/ML opportunities



# Using the Past to Forecast the Future

Leveraging historical data to make better use of land mobile radio (LMR) spectrum

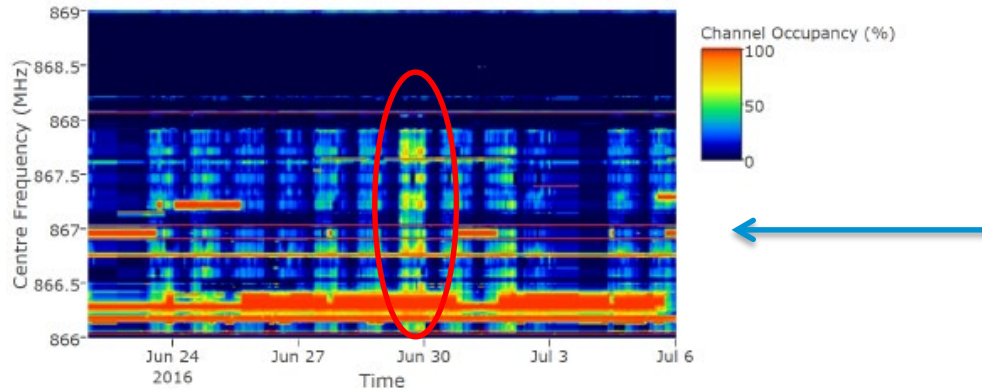


# Spectrum Demand Prediction in LMR Bands

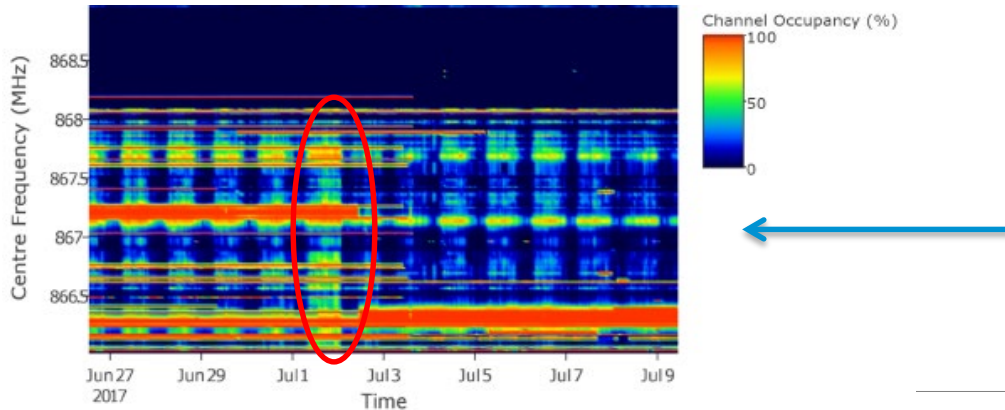




# Spectrum Usage: External Factors

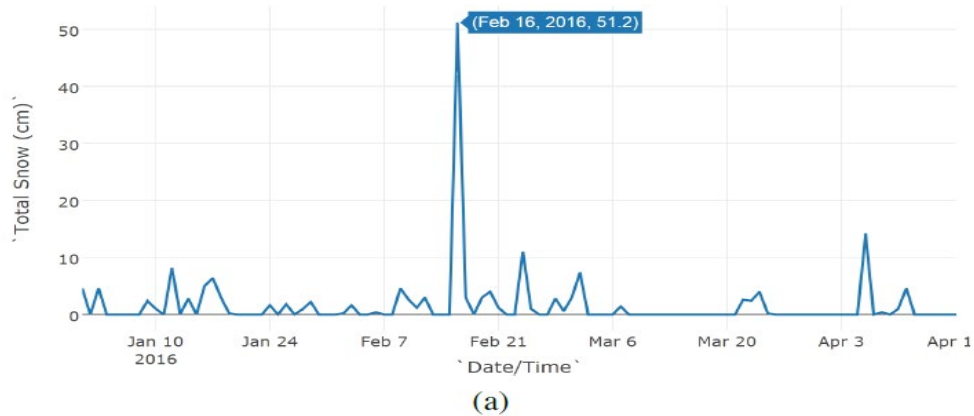


North American leaders' summit in Ottawa

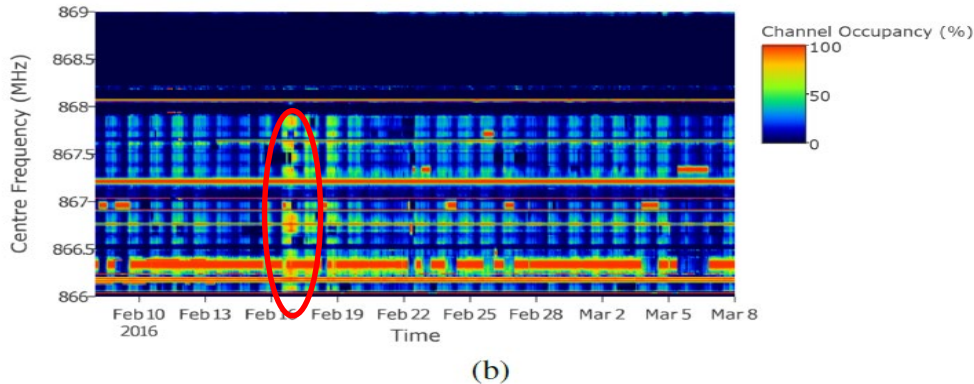


Canada Day 2017

# Spectrum Usage: External Factors



Ottawa's record snowstorm of 2016

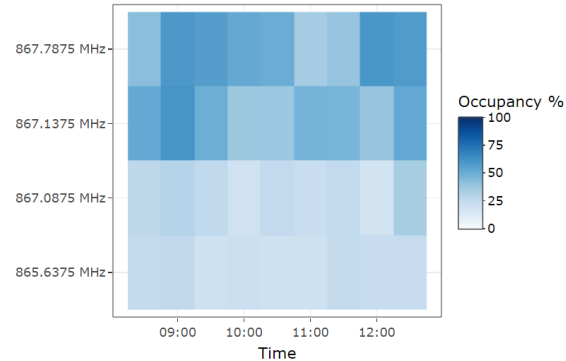




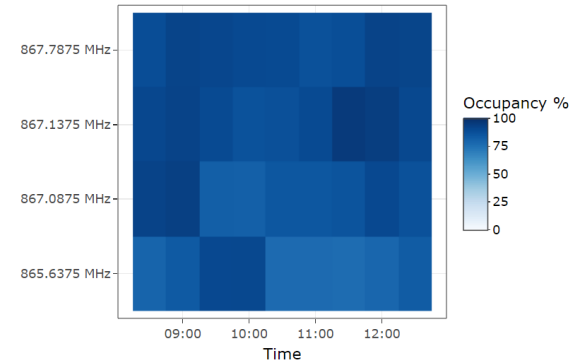
# From Predictive to Prescriptive Analytics

Dynamic sharing of LMR channels with IoT traffic

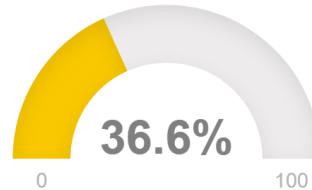
Channel Occupancy Before Sharing (Incumbents-Only)



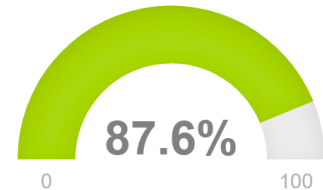
Channel Occupancy After Sharing



Average Occupancy Before Sharing



Average Occupancy After Sharing

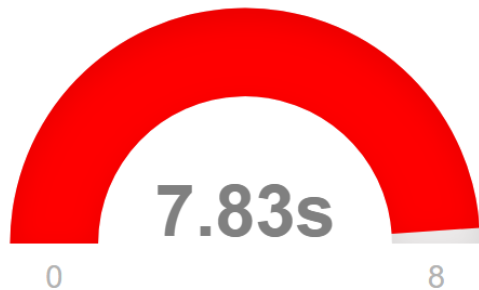


# From Predictive to Prescriptive Analytics

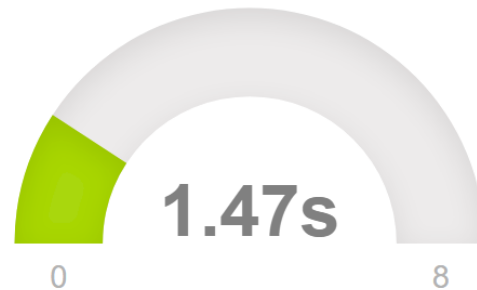
Leveraging ML-predicted LMR usage to reduce interference

## Costs of Sharing

Average Access Delay with Random Sharing



Average Access Delay with ML-Predictive Sharing



# Spectrum Data Beyond LMR

- Commercial mobile bands important to the industry and regulators
- How should we manage the spectrum for 5G and beyond?
- Is sharing viable? If so, under which model(s)?

**Need to understand the “demand behaviour”**

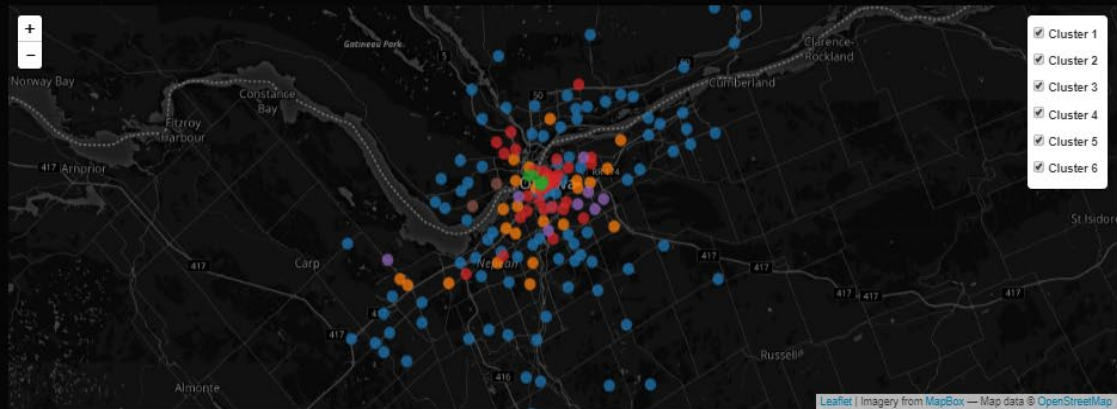
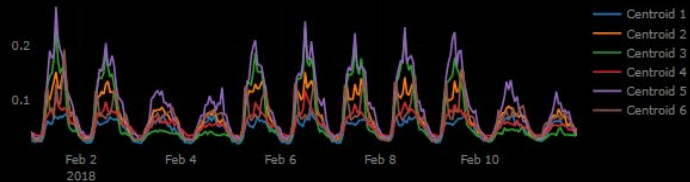
# Characterizing Commercial Mobile Spectrum

- Direct sensing not practical
- Operator data: aggregate demand patterns and ensemble user behaviour
- User data: crowdsourcing applications

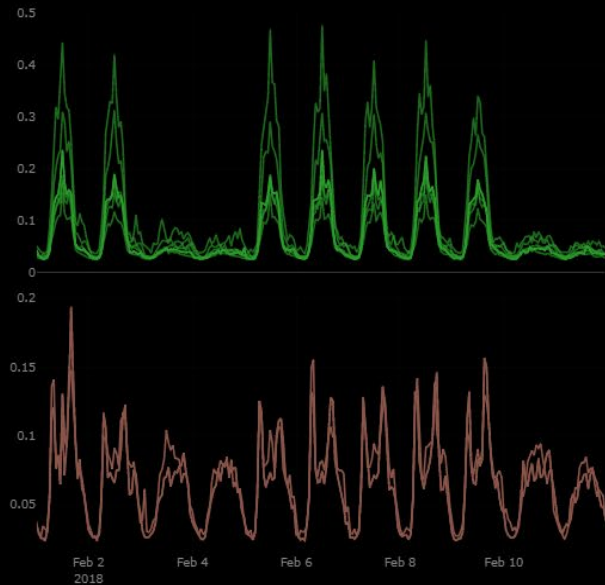
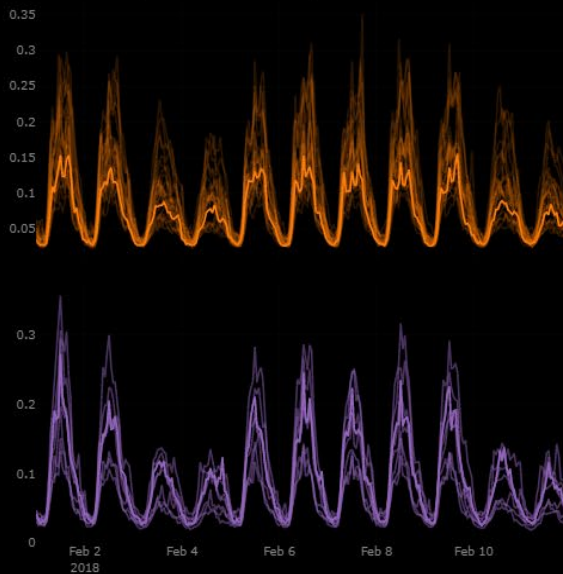
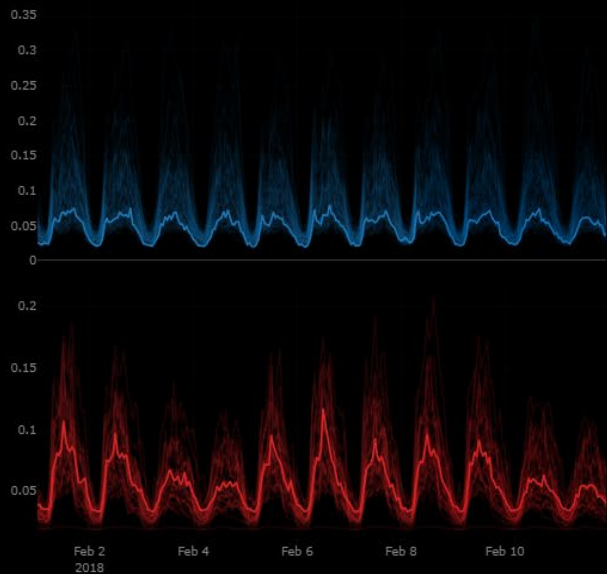
# Ottawa, Canada

## Feb. 2018

Cluster Centroids



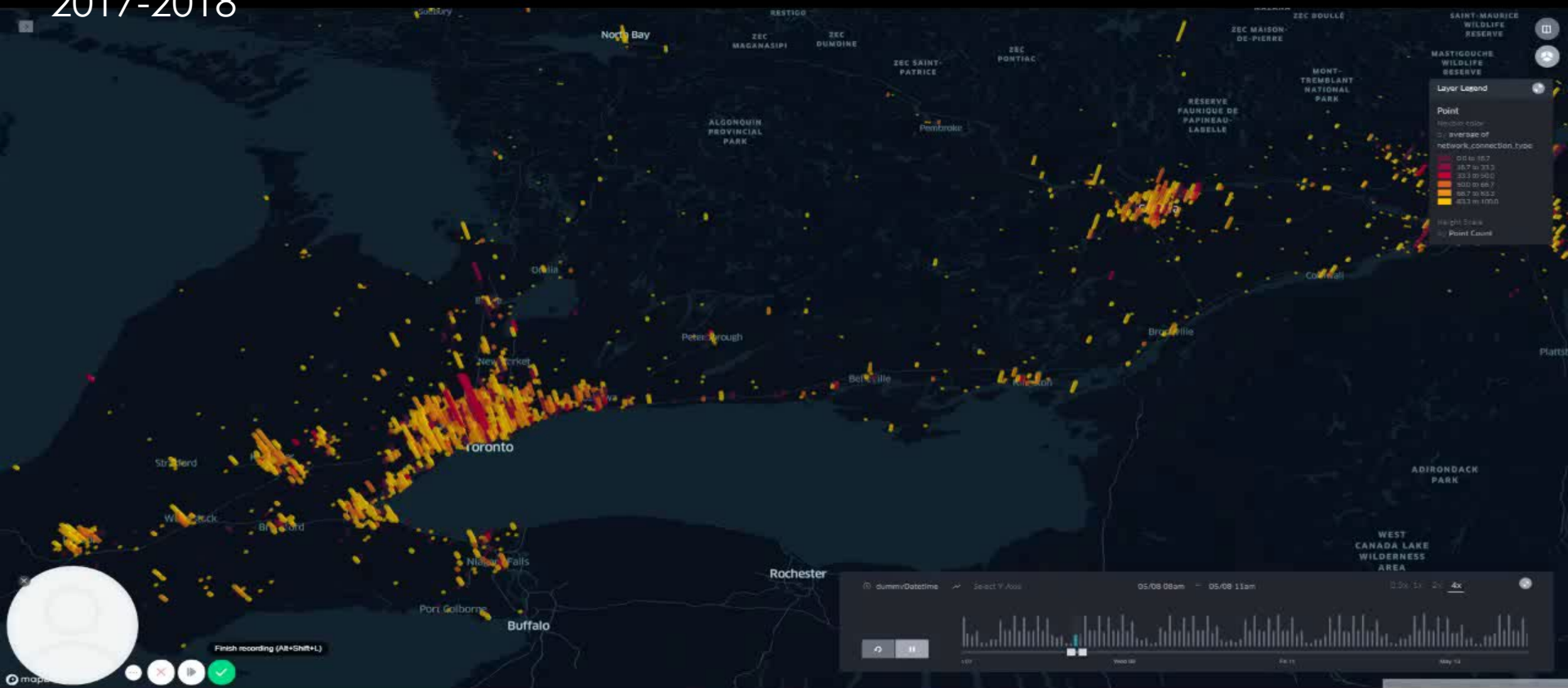
Hourly Resource Block Usage for each Cluster



# Users as Sensors

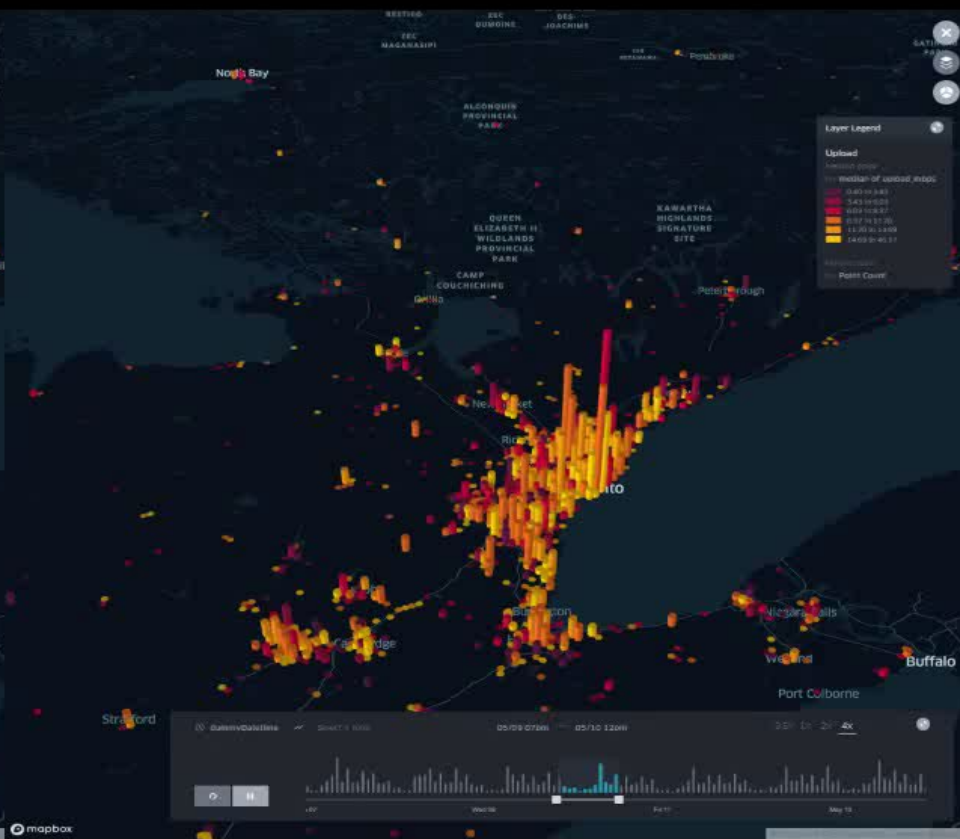
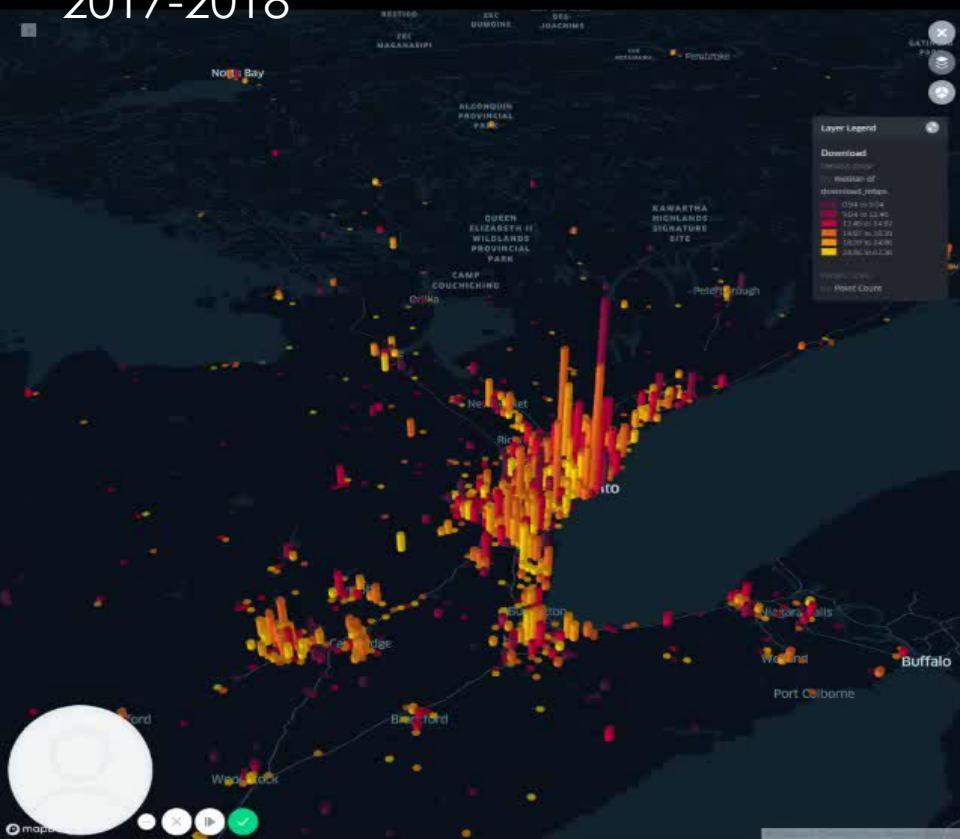
- Crowdsourced mobile spectrum data to complement operator data
  - Models end-user performance
  - Identifies true cell shapes (coverage)
  - Helps detect anomalies

# Ontario, Canada 2017-2018





# Ontario, Canada 2017-2018



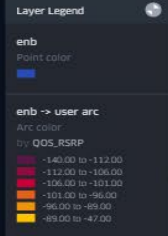
# Ottawa, Canada

## 2018

GATINEAU  
PARK

Gatineau

CHENAULT  
RANGE AND  
PARK  
TRAIL  
CROSSING



# Toronto, Canada

## 2018



# Ottawa, Canada 2018





# Summary

- Spectrum data analytics still in its infancy
- Lots of potential for network optimization, automation, and sharing via ML/AI
- Time and location varying spectrum usage can be leveraged to share spectrum dynamically
- Regulators can use demand and supply insights for dynamic spectrum management

# Summary

- Timely data needed to characterize usage, predict demand, and prescribe actions
- Heterogeneous technologies and services require a diverse data acquisition strategy
- More cooperation needed to define and procure relevant spectrum data

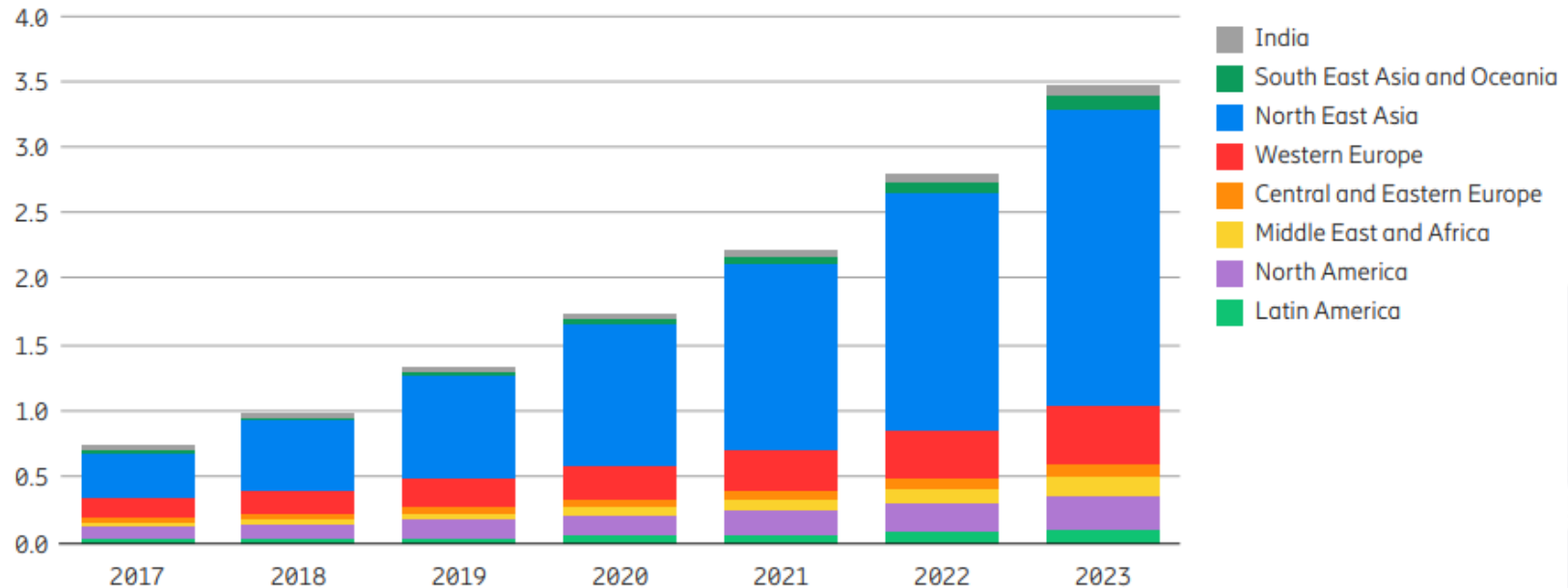
Canada



# Backup Material

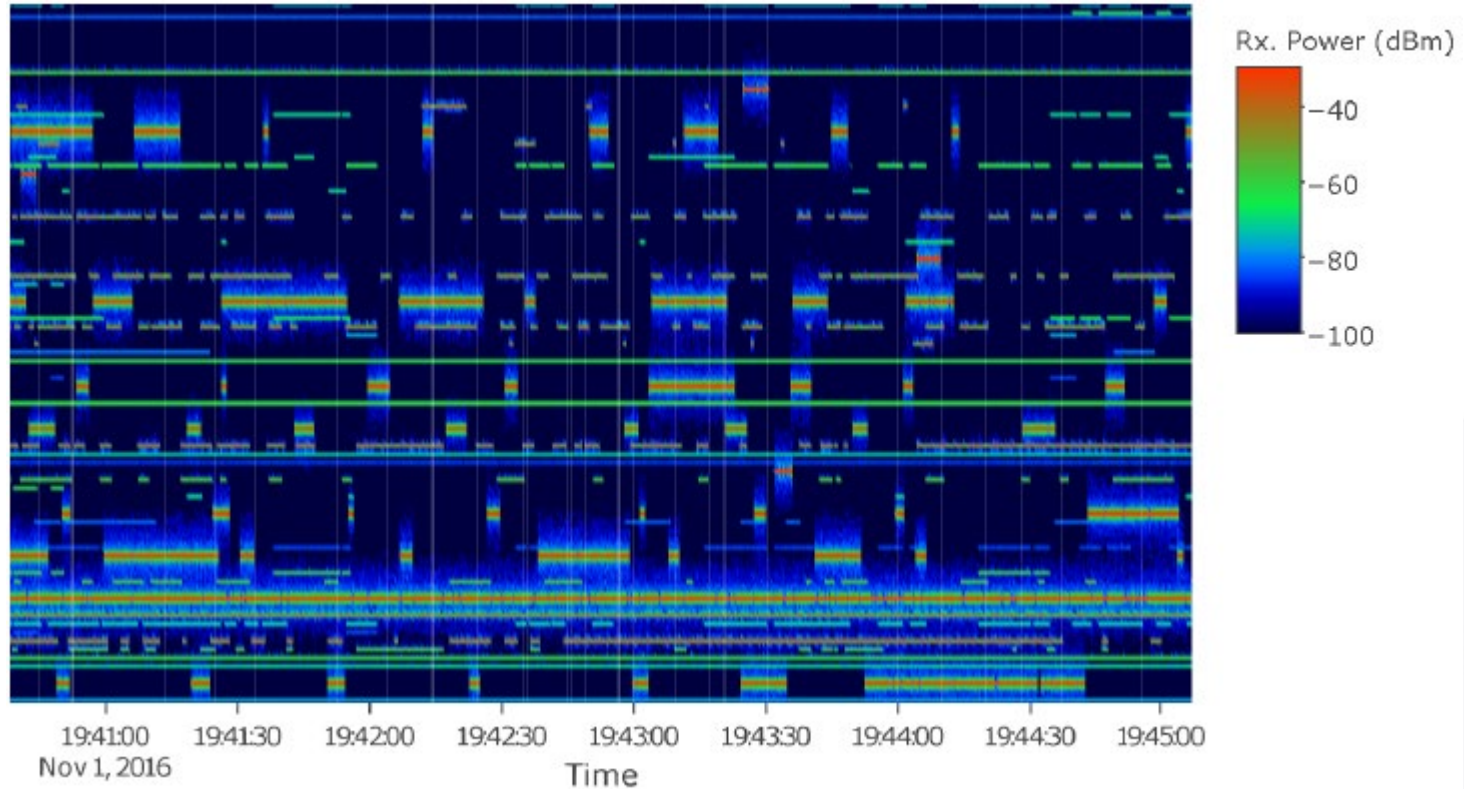
# IoT Growth Forecast

Cellular IoT connections per region (billion)



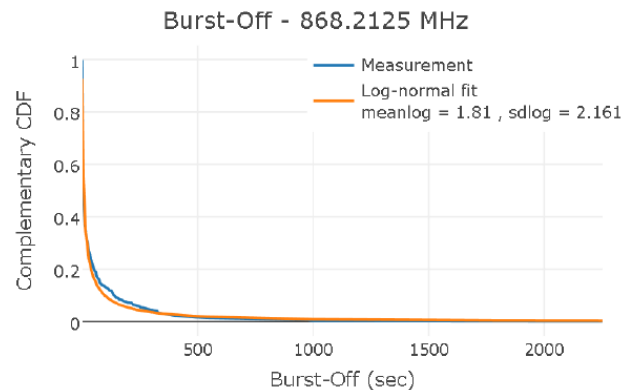
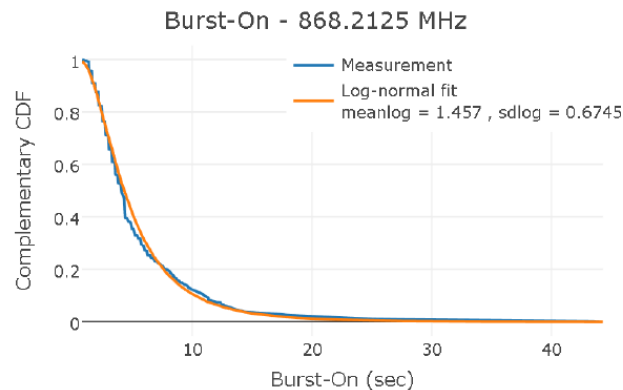
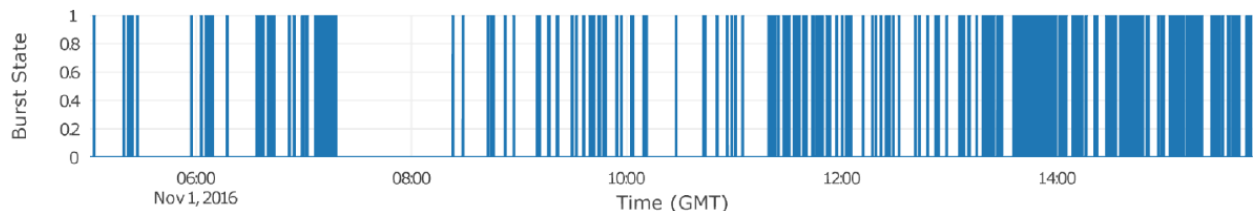
source: Ericsson (Q1 2018)

# Measured Power in Public Safety Land Mobile Radio (866-869 MHz)

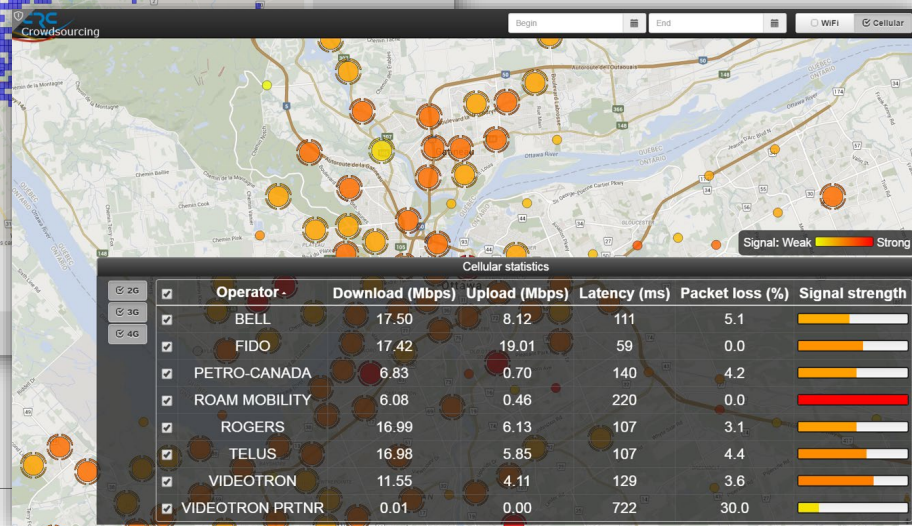
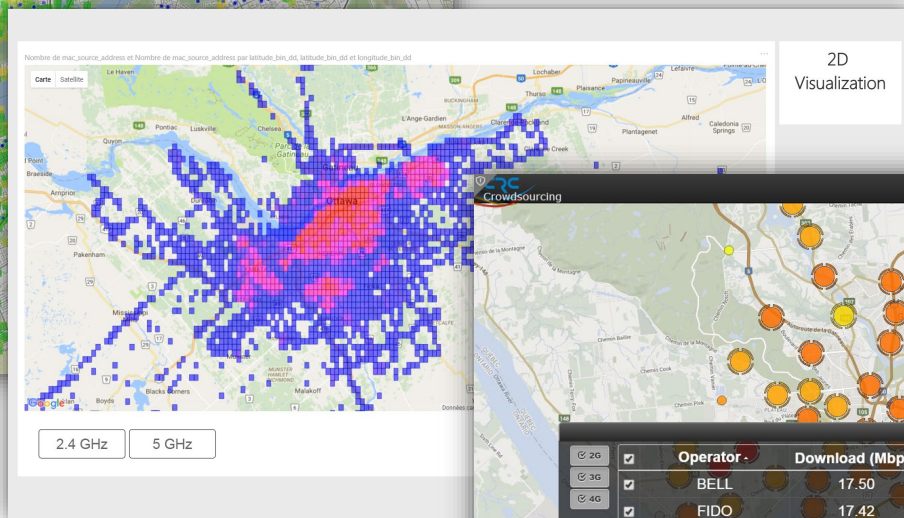
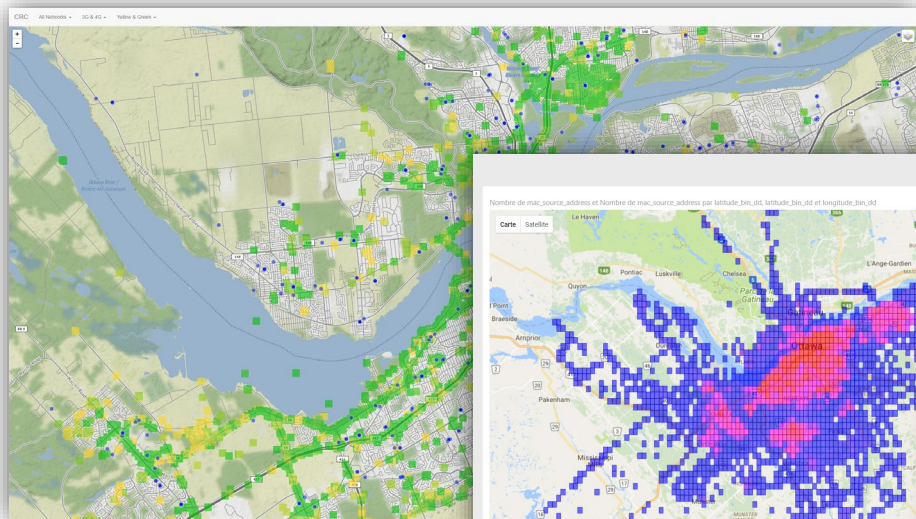


# Land Mobile Radio (LMR) Channel Activity Patterns

- Data-driven traffic modeling
- Quantifying spectrum sharing opportunities



# Crowdsourced Data





# 4G and 5G Coverage Models



# Generic Machine Learning (ML) Pipeline

