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Predicting and Preventing Avoidable Truck Rolls (ATR)

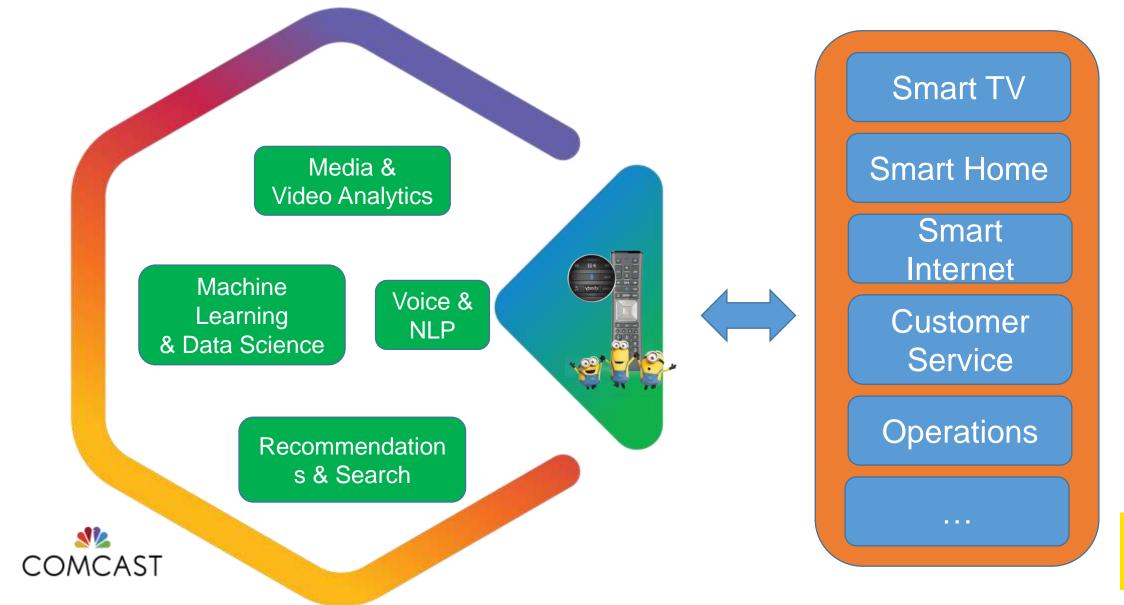
Comcast Applied Al Group







Comcast Applied Artificial Intelligence Group



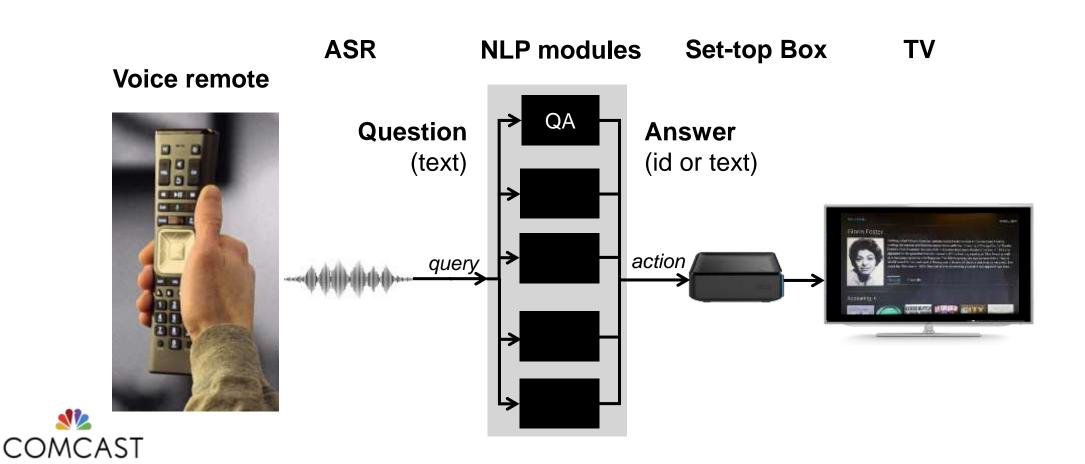


Machine Learning for the Smart TV Voice Interface



X1 Smart TV with Voice

Query: "who plays the oracle in the matrix"





ML for the Smart TV Automatic Content Analysis

Most metadata is at the asset level

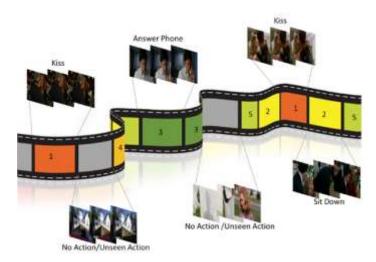
- Genres
- Credits
- Synopsis
- Keywords







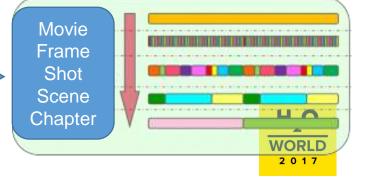




Much mode data Within the asset

- Chapters
- **Moments**
- **Annotations**





Why is this useful?

What are the best moments on TV?

The Tonight Show Starring Jimmy

In-game highlight navigation





Who is in this scene?



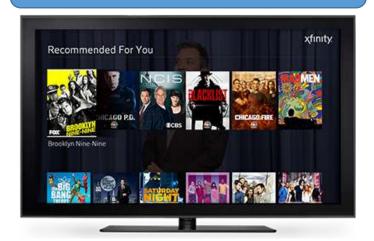
Search & Recommendations





ML Smart TV Personalization Experience

What do you like?





Personalized Recommendations



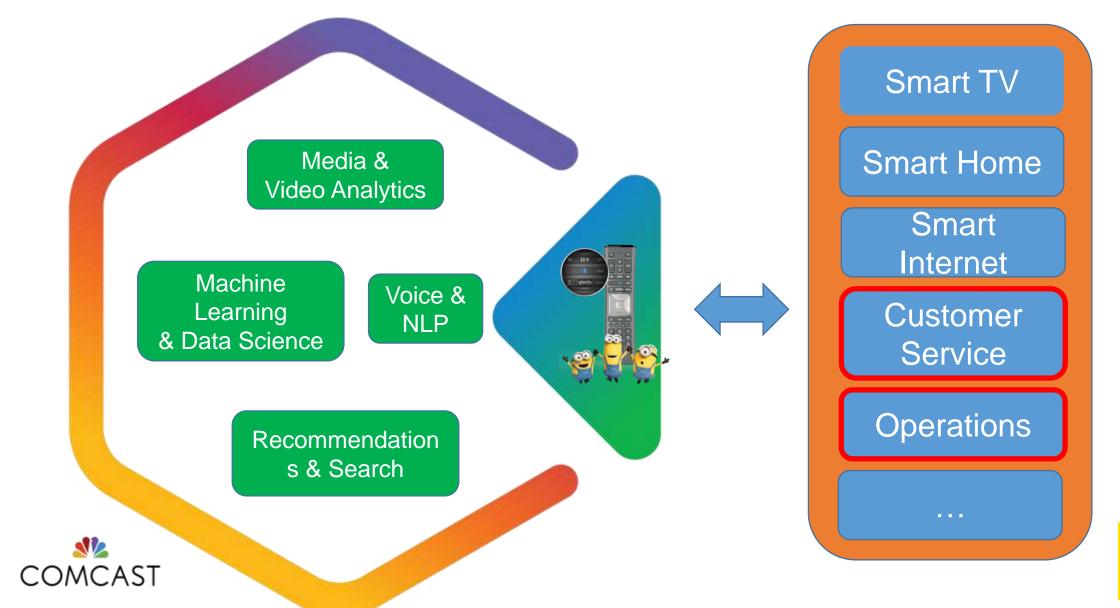








Predicting and Preventing (ATR)





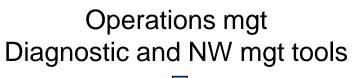
Customer Care + Al

Additional data sources: several GB/day



System Errors

XRE errors



Device errors



Customer Agent

• Call for video quality issues



GoalBetter customer
Experience



Truck Roll

Truck dispatch





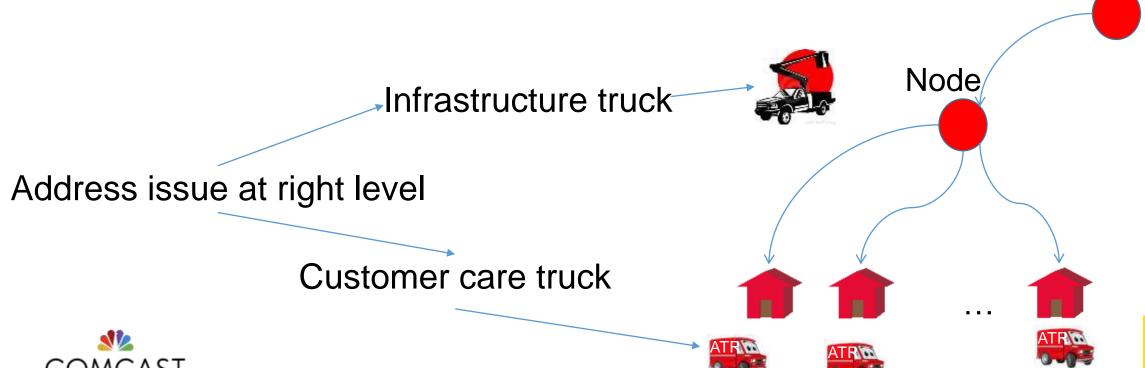


Avoidable Truck Roll (ATR)

Problem

Predict and prevent ATRs

don't roll a truck to the home when deeper network issues exist.







ATR and Customer Experience

ATR

A large percentage of customer issues for which truck rolls are ordered end up being unnecessary

Better customer experience

Reducing truck rolls increases customer experience and improves NW service reliability

Lack of insight

Into what's triggering excessive truck rolls. Service departments struggle to reduce truck rolls because they don't understand the root causes and how to address them





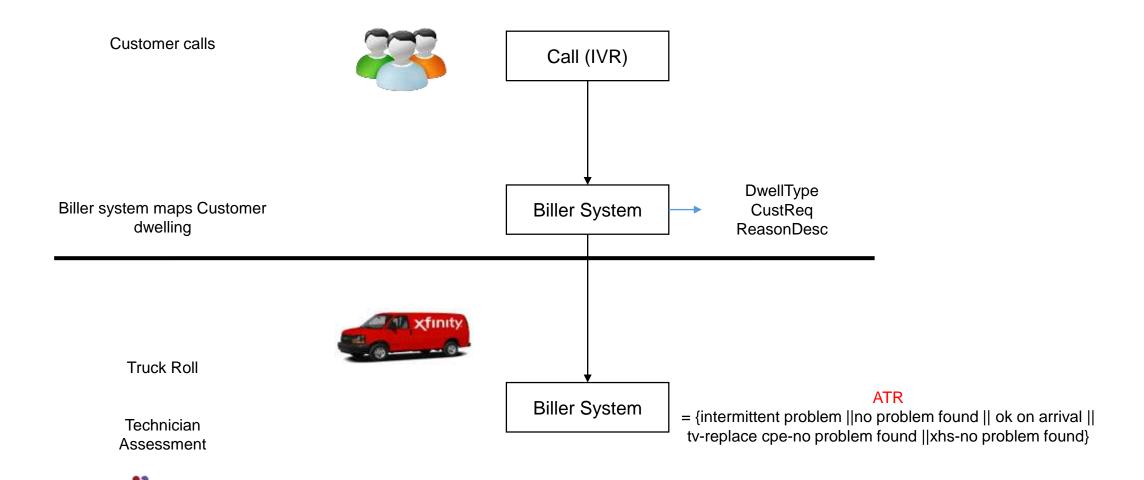
First Model Nov 28 2016

Truck Roll info + NW health





Following the Truck Roll Flow



COMCAST



Assessing Network Health

NW Health issues Node status messages

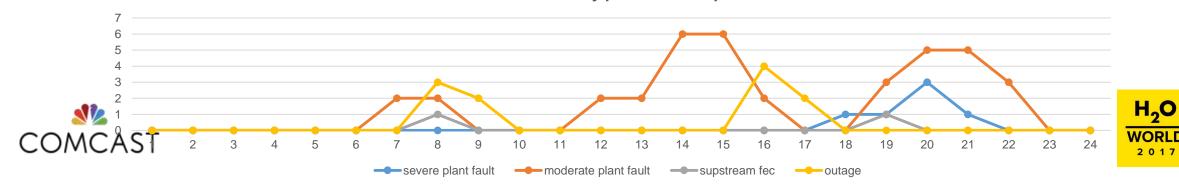
nhealth_etypedesc, e.g. moderate plant fault, upstream fec, severe plant fault, outage, ...

nhealth_etype, nhealth_stateid, nhealth_devicetype Assessing NW health



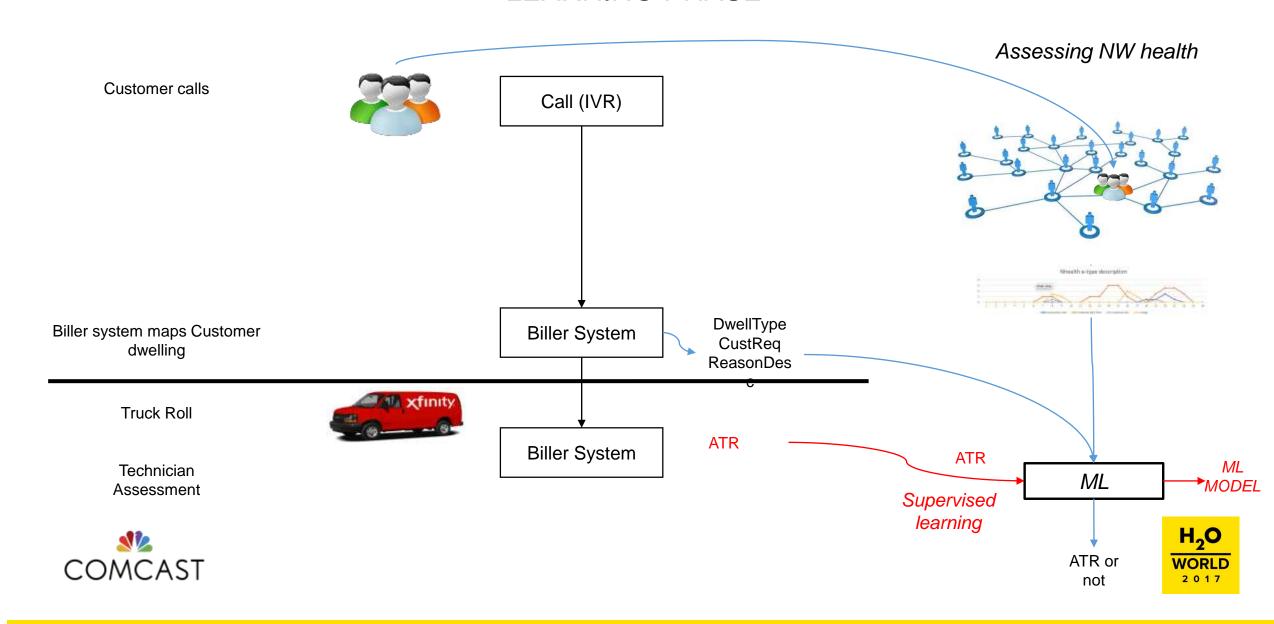
These messages are represented in Time-Series

Nhealth e-type description



Our Approach to Predict Avoidable Truck Rolls

LEARNING PHASE



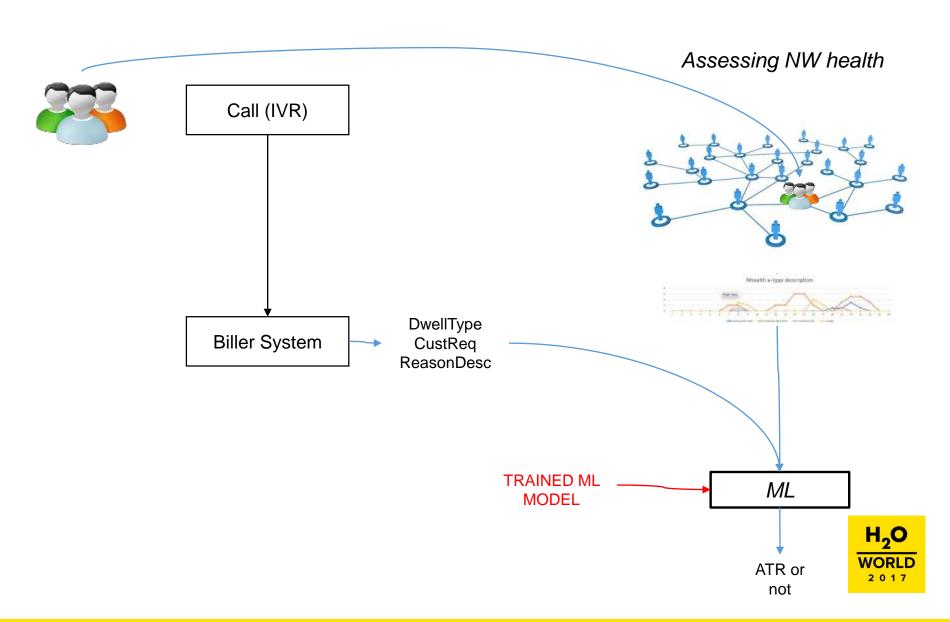
Our Approach to Predict Avoidable Truck Rolls

PREDICTION PHASE

Customer calls

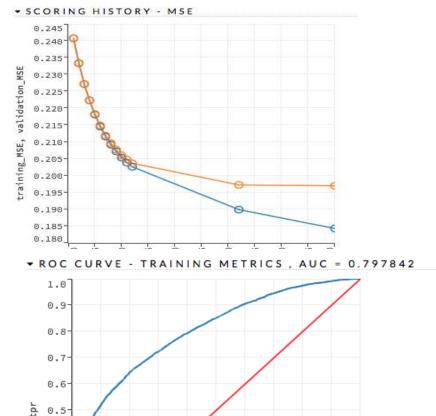
Biller system maps Customer

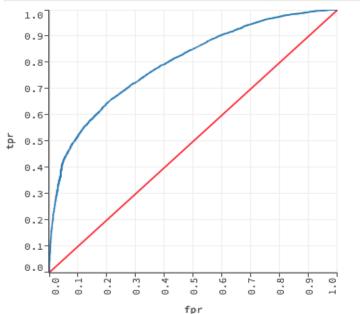
dwelling





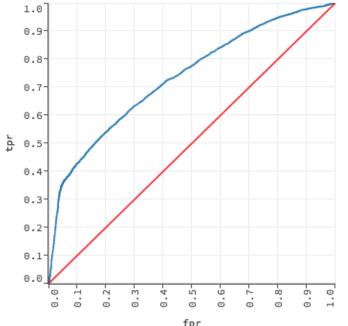
Simulation Results: 67% accuracy





metric	threshold	value	idx
max f1	0.3377	0.6943	286
max f2	0.0712	0.8337	396
max f0point5	0.5790	0.6879	136
max accuracy	0.5079	0.6700	172
max precision	0.8617	0.8947	44
max absolute_MCC	0.7054	0.3829	89
max min_per_class_accuracy	0.4403	0.6617	214









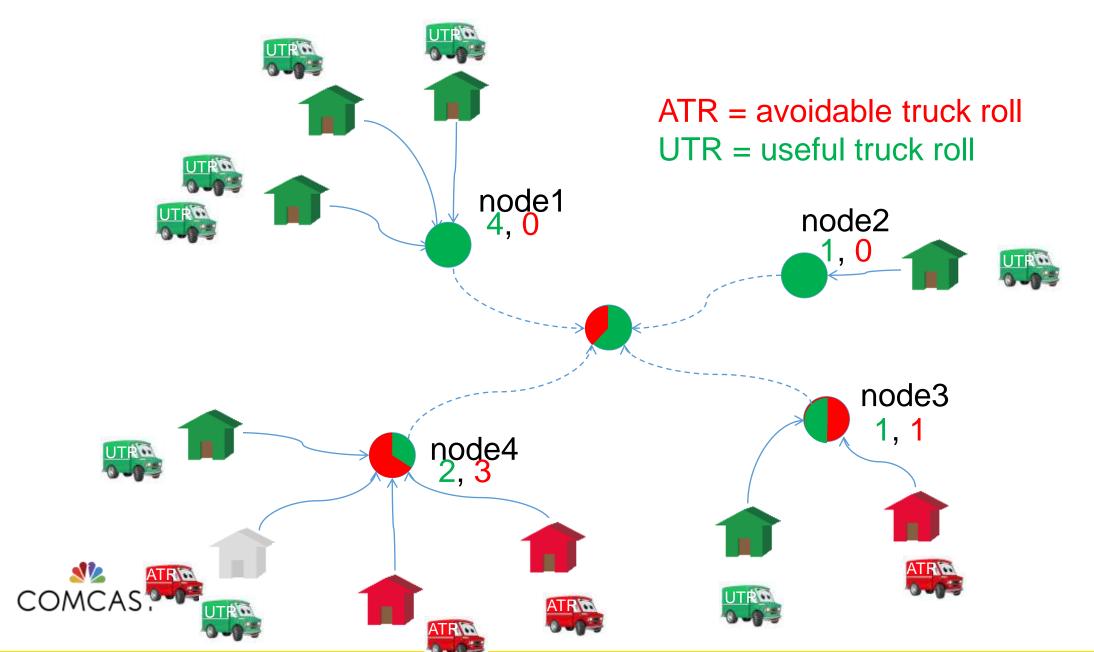
Second Model Dec 2 2016

1st Model + Addition of NW Topology



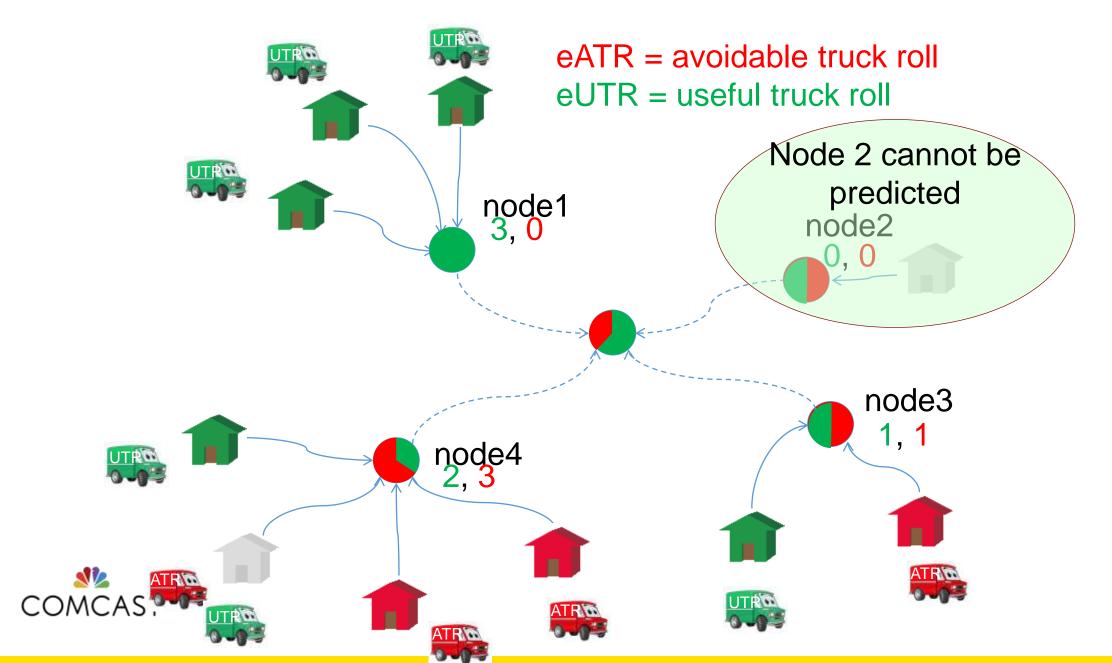


Addition of Neighborhood Topology



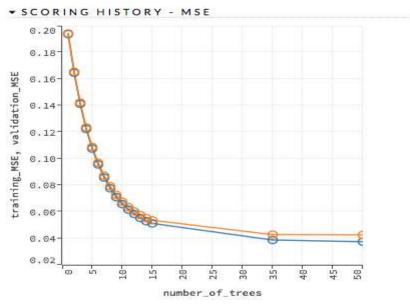


Estimating the Neighborhood Topology in Real-Time



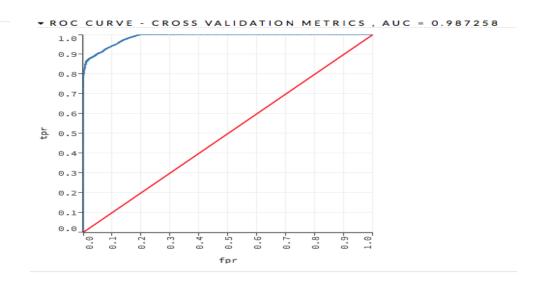


Simulation Results: 93% Accuracy



metric	threshold	value	idx
max f1	0.4828	0.9490	265
max f2	0.1467	0.9783	320
max f0point5	0.8424	0.9639	110
max accuracy	0.5106	0.9325	254
max precision	0.9922	1.0	0
max absolute_MCC	0.4828	0.8536	265
max min_per_class_accuracy	0.5964	0.9242	209

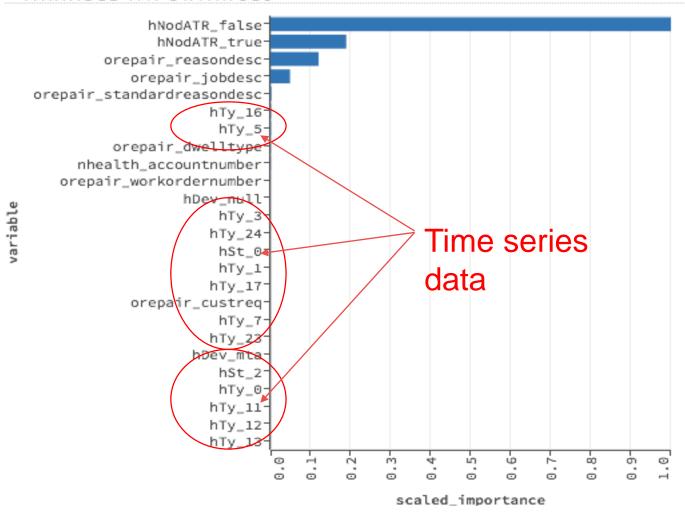






Modest Contribution of Time Series

▼ VARIABLE IMPORTANCES



Not worth the hassle

Drop them In next model

and replace them by NW features





Third Model May 15 2017

Extended definition of ATR In depth revisited model No more time series





Following the Call Center Flow

Customer calls

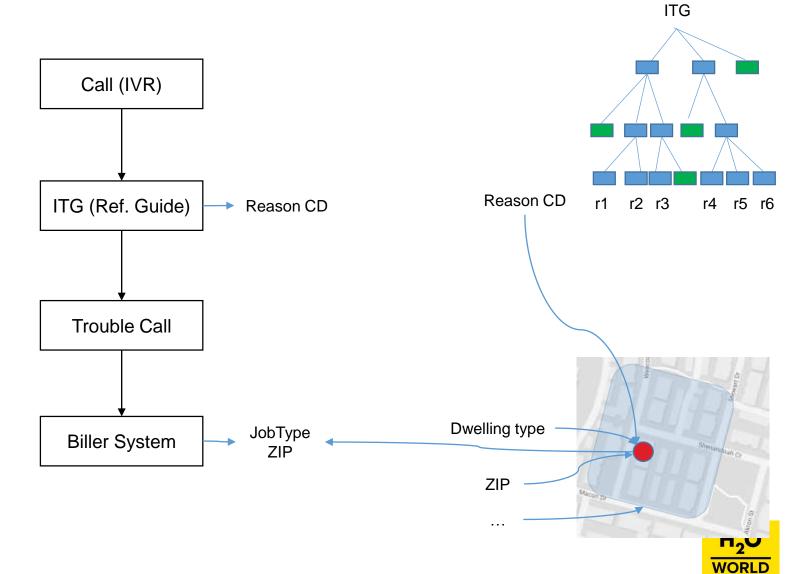
3

Agent follows Interactive Trouble shooting Guide



Failure, create Ticket

Biller system maps ReasonCD to Customer dwelling, ZIP, ...

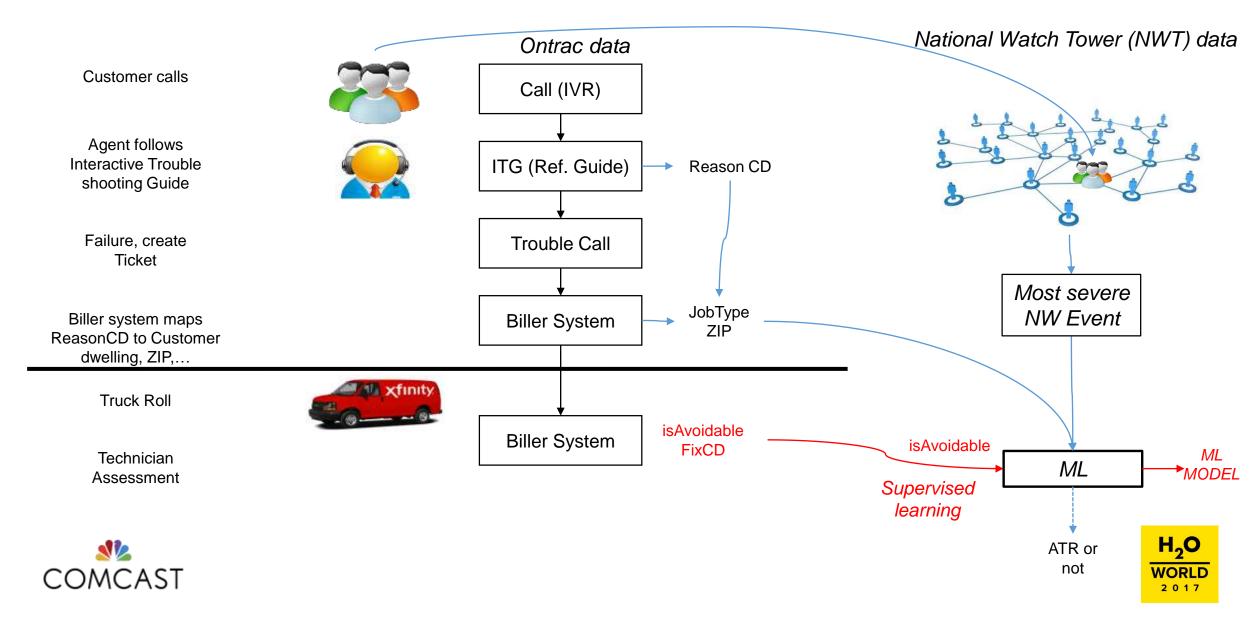


2017



Our Approach to Predict Avoidable Truck Rolls

LEARNING PHASE



Our Approach to Predict Avoidable Truck Rolls

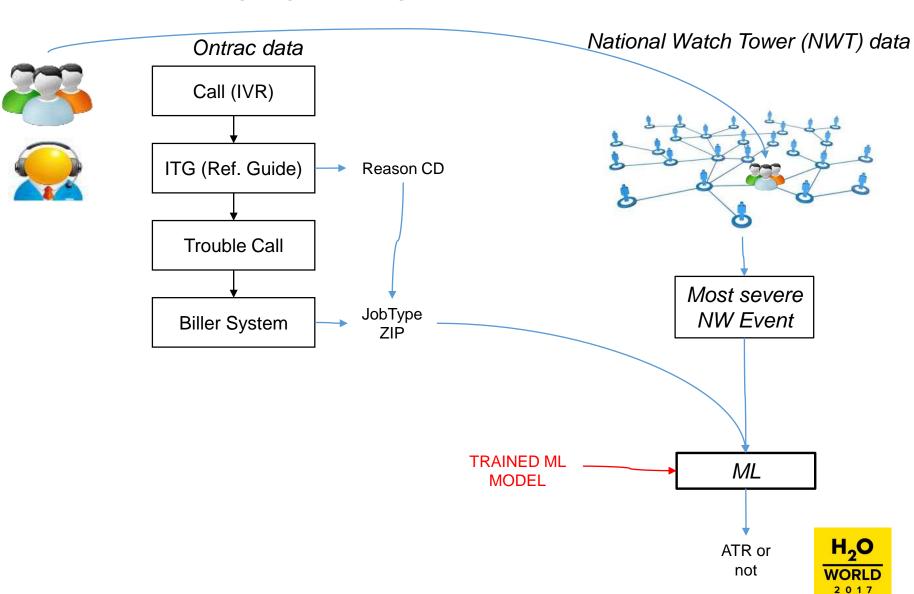
PREDICTION PHASE

Customer calls

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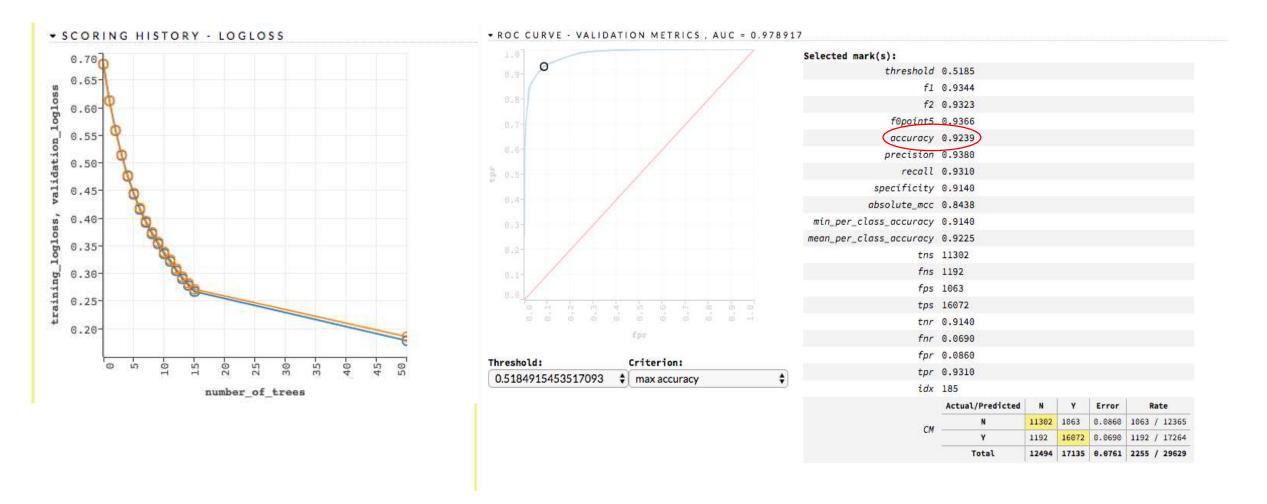
> Failure, create Ticket

Biller system maps ReasonCD to Customer dwelling, ZIP, ...





Simulation Results: 92% Accuracy







Initial Deployment Results: 90% Accuracy

(when available)







Conclusion

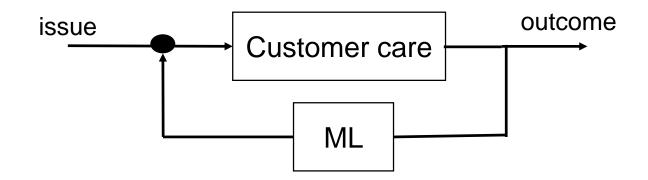
Good proof of concept on ATR detection

Limitation of the method

This method only works once the customer has placed a call
The method needs more investigation to define the root cause of the ATR

A proactive method preempting calls will have more impact on customer experience

We enter in a new era of customer care by developing a method based on Reinforcement Learning









COMCAST : LABS :



