



Understanding automated wait time data

Checkpoint 2 analysis



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Topics:

- How accurate are the predicted wait times?
- What are the drivers of high predicted wait time?
- Conclusions, recommendations and considerations



**HOW ACCURATE ARE THE
PREDICTED WAIT TIMES?**

Findings

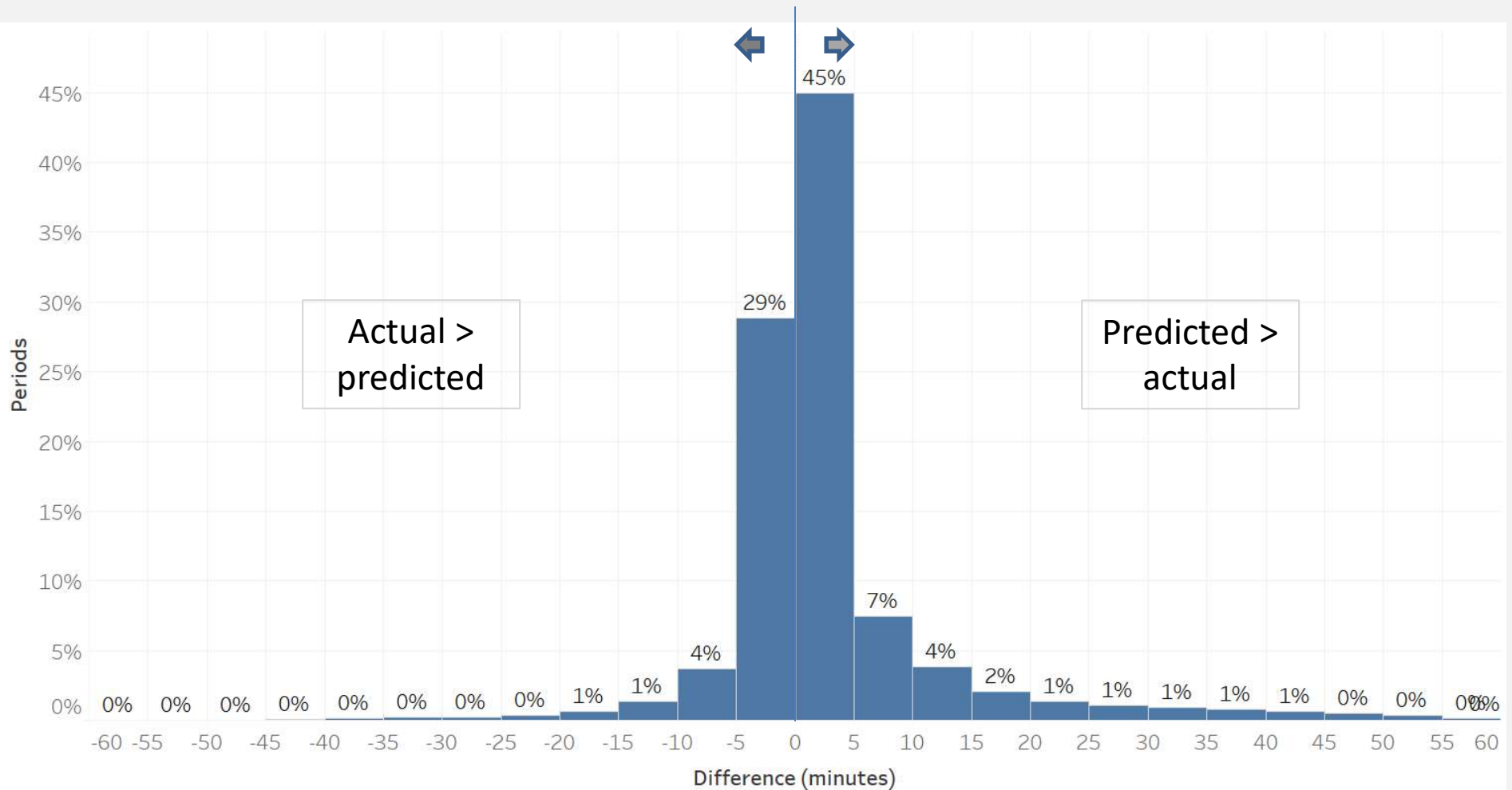
Wait times were within **five** minutes of prediction **74%** of time*

*System awaiting final configuration (XOVIS) and user acceptance testing/validation (Port of Seattle)

Analysis

- Analyzed data since Jan 2018 for Chkpt. 2
- Over 2,500 hours/152,000 minutes evaluated
- Compared actual and predicted times
- Tracked frequency of high line wait times and large differences between predicted and actual

Predicted time generally > actual wait,
most often within five minutes



- 74% within (+/-) 5 minutes
- 85% within (+/-) 10 minutes
- 90% within (+/-) 15 minutes

Predicted vs. actual times

A photograph of a busy airport security checkpoint. A large crowd of people is waiting in a line, separated by metal stanchions. In the foreground, a man in a teal shirt is seen from behind, looking towards the line. Above the queue, a green sign reads "END OF THE LINE". To its right, a black sign features the text "A Pre✓" and a red sign with a white upward-pointing arrow. The scene is brightly lit with overhead lights, and the floor is a light-colored tile. The overall atmosphere suggests a long wait time.

**WHAT ARE THE DRIVERS OF HIGH
PREDICTED WAIT TIMES?**

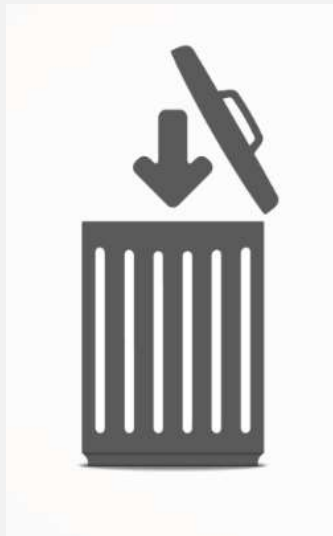
High predicted wait times or big differences between predicted and actual waits largely the result of

- System configuration issues
- Premium lane demand
- TSA staffing changes (lanes opening/closing; canines)
- Load balancing

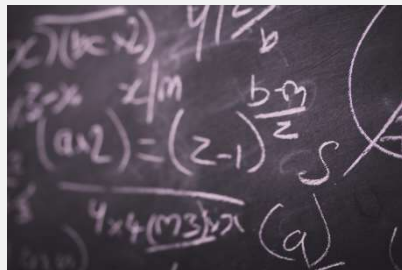
Predictive Analysis

Identifying patterns to predict the probability of something happening.

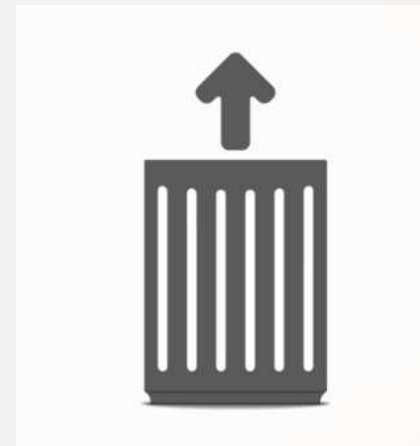
Data



Models



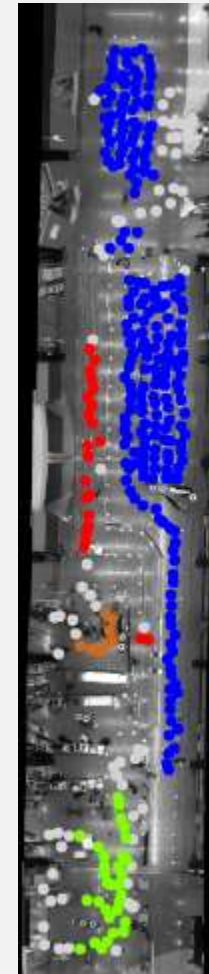
Prediction



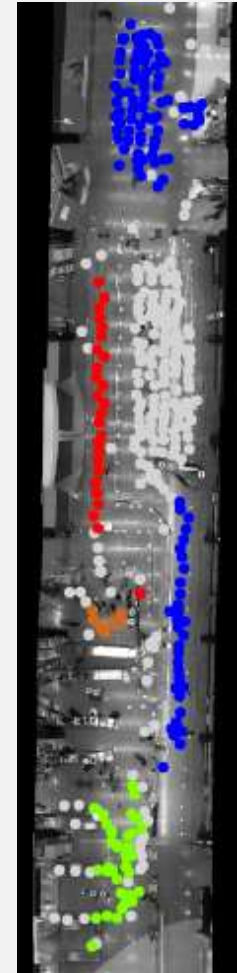
Configuration issues



@ 2:25pm



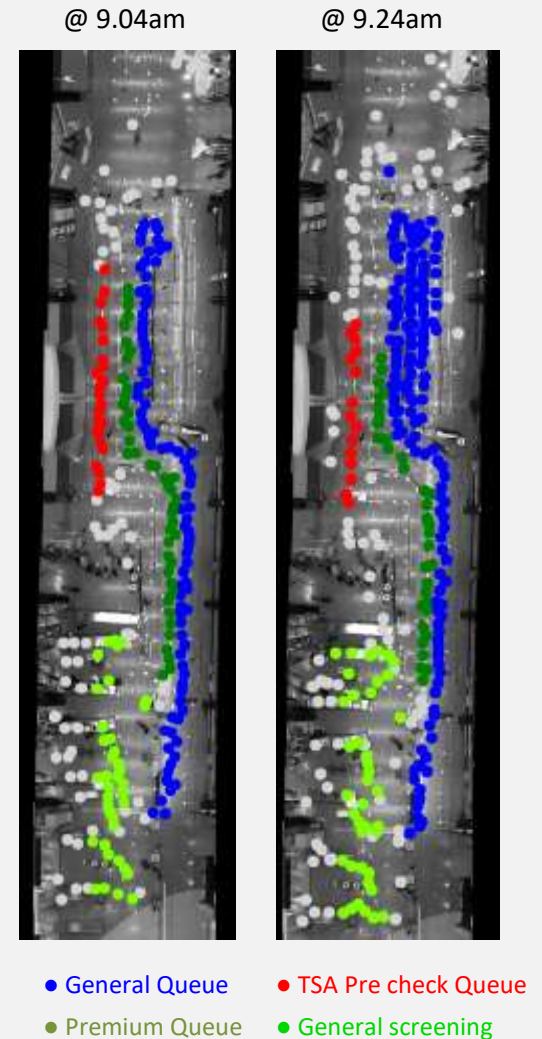
@ 2:27pm



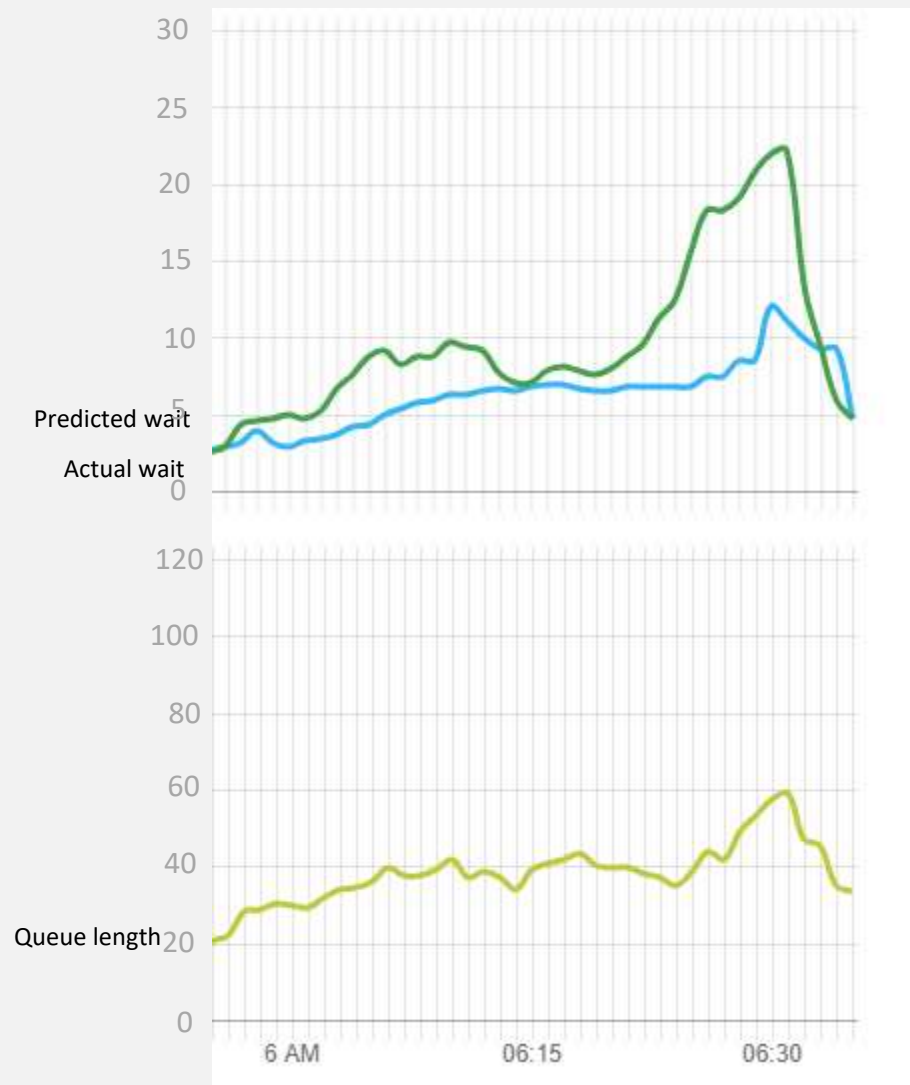
- General Queue
- TSA Pre check Queue
- Premium Queue
- General screening
- Premium Screening

Premium queue

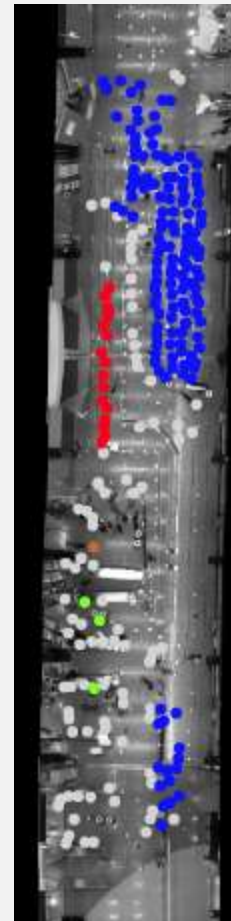
- Prioritization
- Premium demand drives up general screening wait times
- XOVIS predicted waits reflect scenario well
- Lane classification accuracy has opportunity for improvement



TSA canine



6:31am



6:35am



● General Queue

● Premium Queue

● TSA Pre check Queue

● General screening

● Premium Screening

Slide 13

SD1

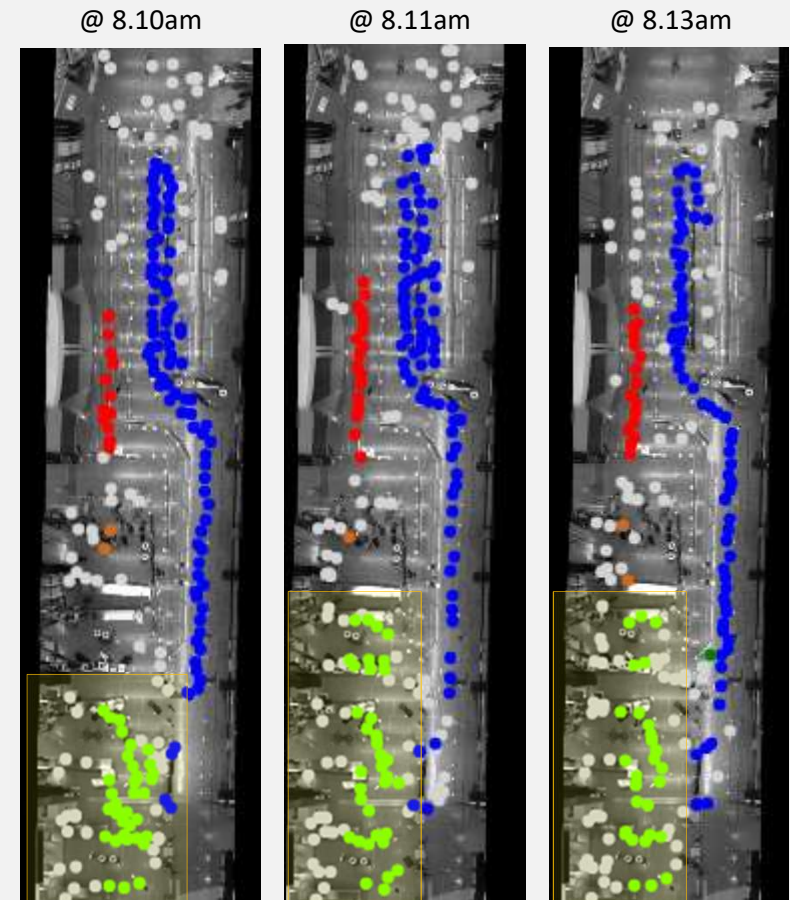
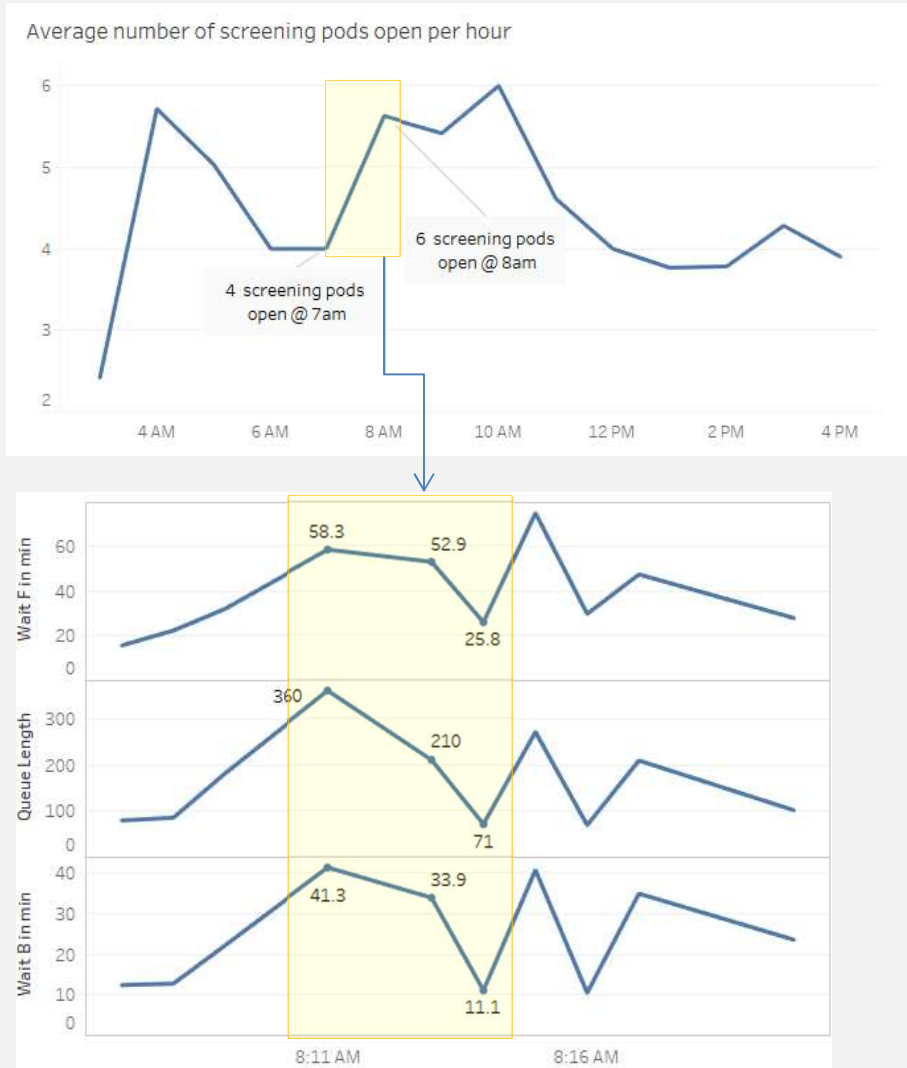
Decided against lables on Xovis images

Was too cluttered

See talking points

Shasha, Daniela, 5/18/2018

TSA staffing changes

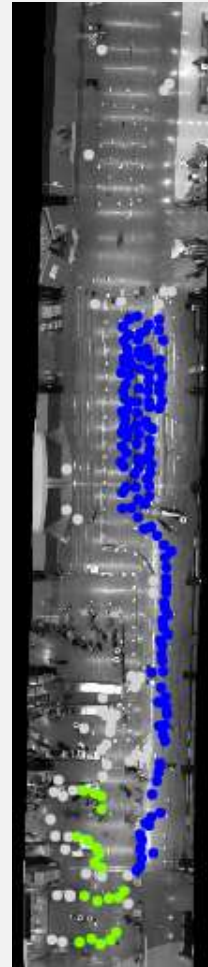


- General Queue
- Premium Queue
- TSA Pre check Queue
- General screening
- Premium Screening

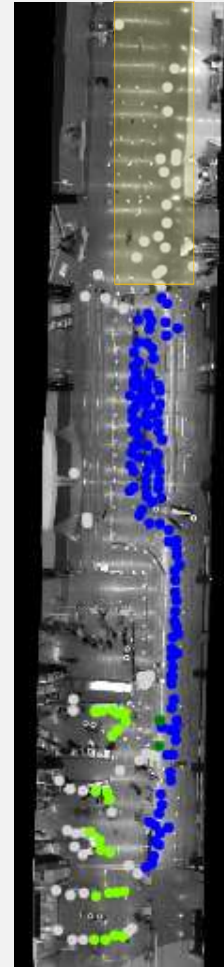
Load balancing



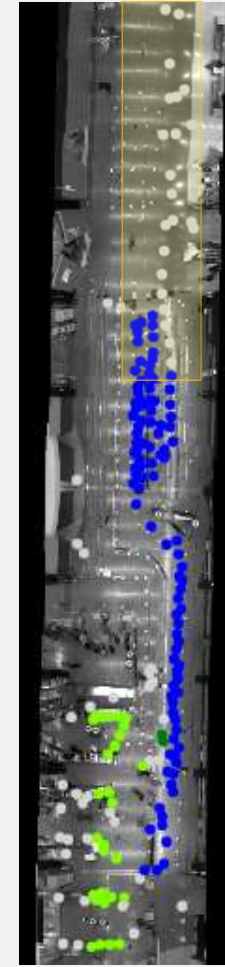
@ 3.47pm



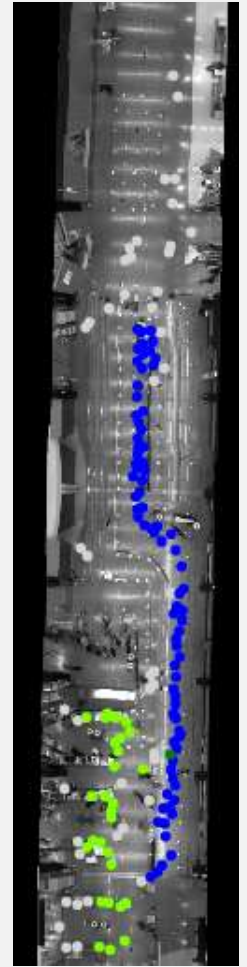
@ 3.48pm



@ 3.49pm



@ 3.50pm



• General Queue

• Premium Queue

• General screening

Busy periods = more occurrences

Large differences in predicted vs. actual wait times occur most often during high demand periods.

Hour	Day of Week						
	Sun	Mon	Tue	Wed	Thu	Fri	Sat
4							
5							
6							
7							
8							
9							
10							
11							
12							
13							
14							
15							
16							
17							
18							
19							
20							
21							
22							
23							

Source: analysis by Port Business Intelligence

CONCLUSIONS, RECOMMENDATIONS AND ITEMS FOR CONSIDERATION

Conclusions

- Technology is largely working as intended
- Configuration adjustments will further improve predictive and actual wait measures
- Certain types of checkpoint events (e.g. TSA staffing level changes, load balancing) will cause spikes/volatility in wait times.

Recommendations

- Have XOVIS proceed with final system configuration with Port input and have BI rerun analysis
- Have BI validate actual waits as part of UAT process.

Considerations

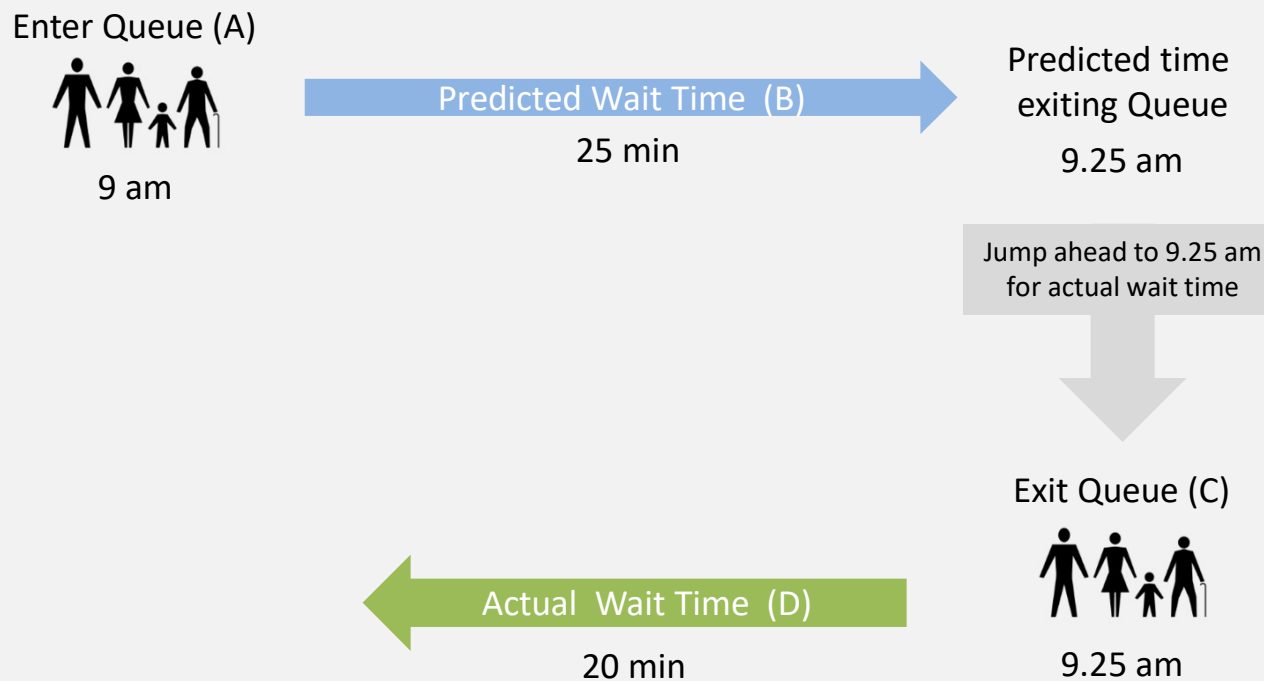
- XOVIS deployment is designed to promote self-load balancing by informing travelers of wait time – is the Port's deployment able to fully leverage this function?

Thank you

Appendix

Methodology

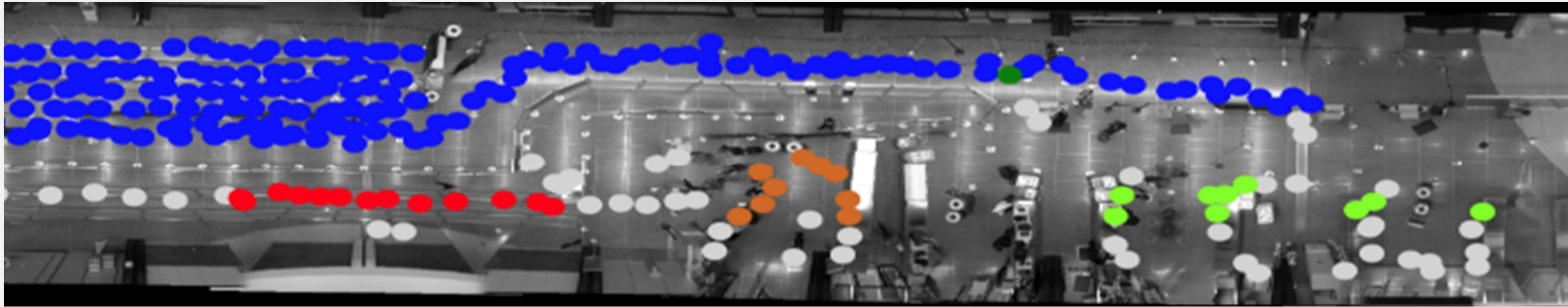
$$\text{Forecast Variance} = \text{Predicted (B)} - \text{Actual (D)}$$



Enter Queue (A)



Predicted Wait Time (B)



Exit Queue (C)



Actual Wait Time (D)



$A + B = \text{Exiting Time Period to Observe}$

$\text{Actual (D)} - \text{Predicted (B)} = \text{Forecast Variance}$

Enter Queue (A)



9 am

Predicted Wait Time (B)

25 min

Predicted time
exiting Queue

9.25 am

Jump ahead to 9.25 am
to get actual wait time

9.25 am

Exit Queue (C)



20 min

Actual Wait Time (D)

Forecast Variance = Actual (D) - Predicted (B)

Data

- Source_name = WT_CP2_new
- Configname= General Queue
- WaitB = actual time waited
- WaitF = predicted wait time
- Data filtered to 4am to 11pm in Tableau

Clarifications

- Account for empty line/staff? (Validate in client)
- Include upper limit in histogram? Up to 15min
- Considerations for using pax and processing rate, counters for further analysis