
Impact of EOBT Uncertainty on Airport Surface Congestion Management

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Sponsor: Chris Dorbian, FAA Office of Environment & Energy



NASA ATD-2 Industry Workshop

4-5 September 2019



Motivation



- **Airport surface congestion leads to increased taxi time, fuel burn & environmental impacts**



- **Advanced automation systems are under development to reduce surface congestion**
 - **FAA Terminal Flight Data Manager (TFDM)**
 - **NASA Airspace Technology Demonstrator-2 (ATD-2)**



- **Effectiveness of systems depend on algorithm design and accuracy of key input data, e.g., Earliest Off-Block Time (EOBT)**

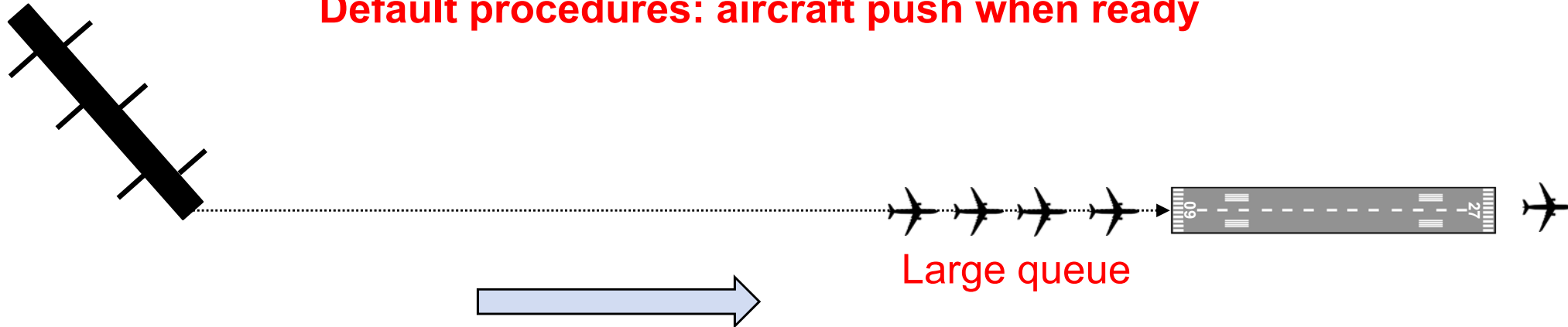
Need for analysis to understand relationship between automation system benefits and EOBT accuracy to inform future algorithm design and airline investments



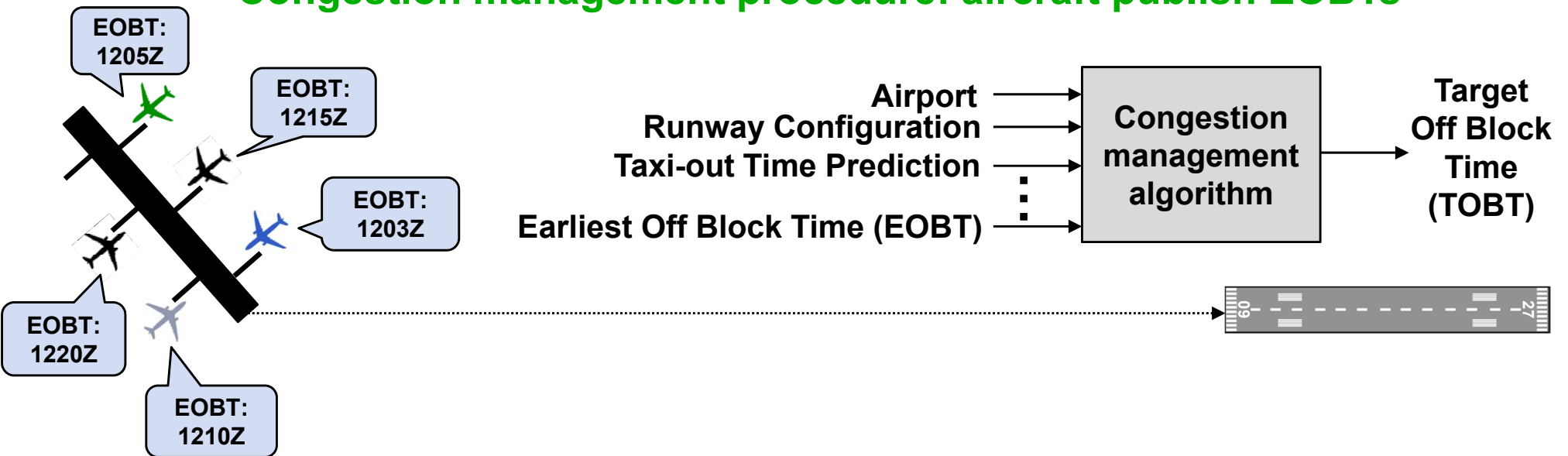
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Default procedures: aircraft push when ready



Congestion management procedure: aircraft publish EOBTs

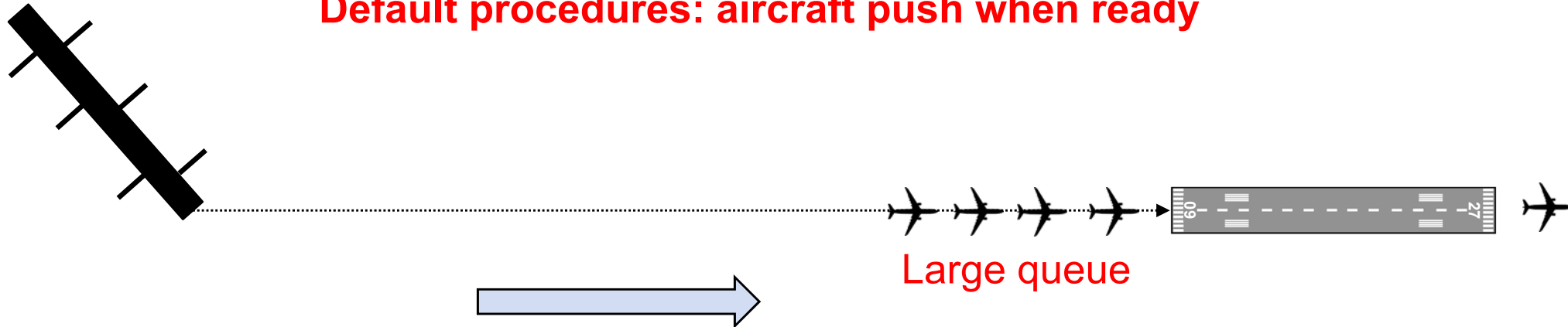




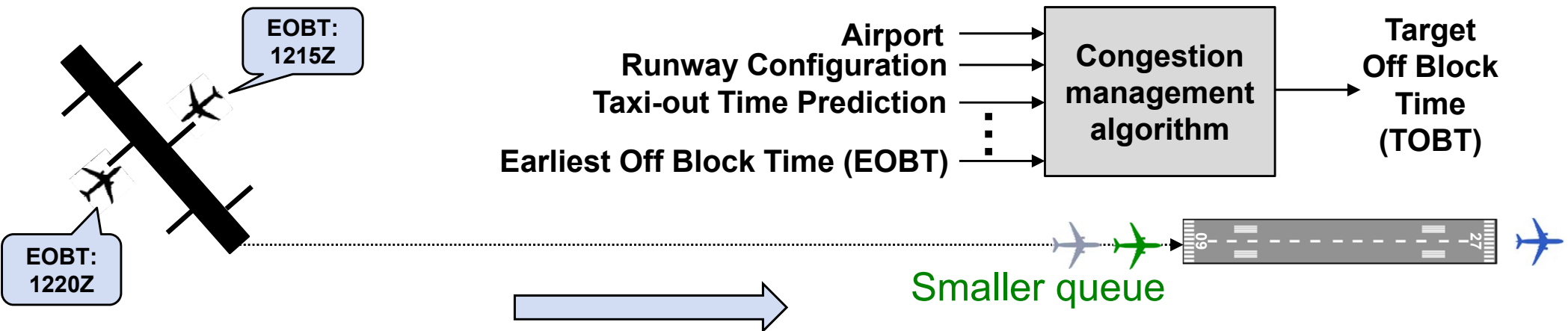
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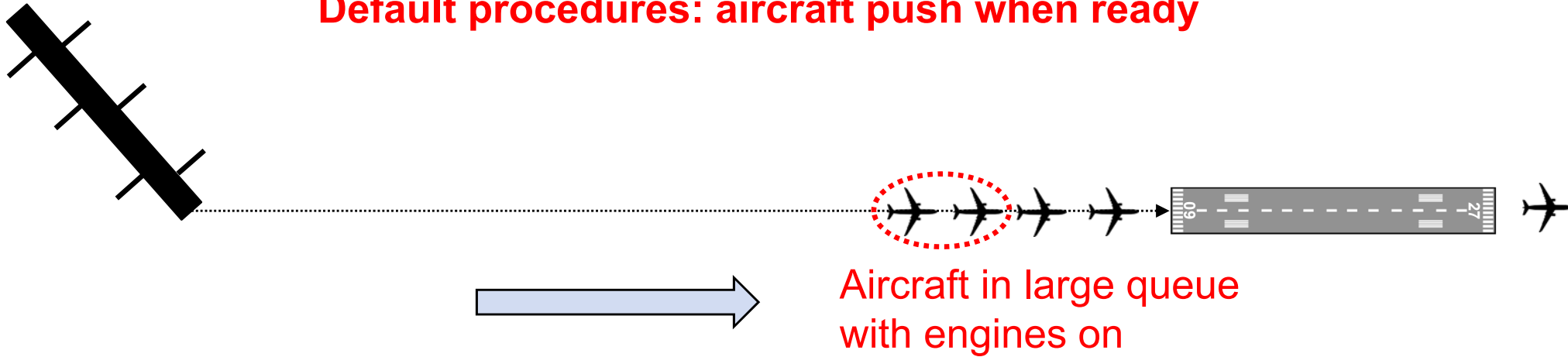




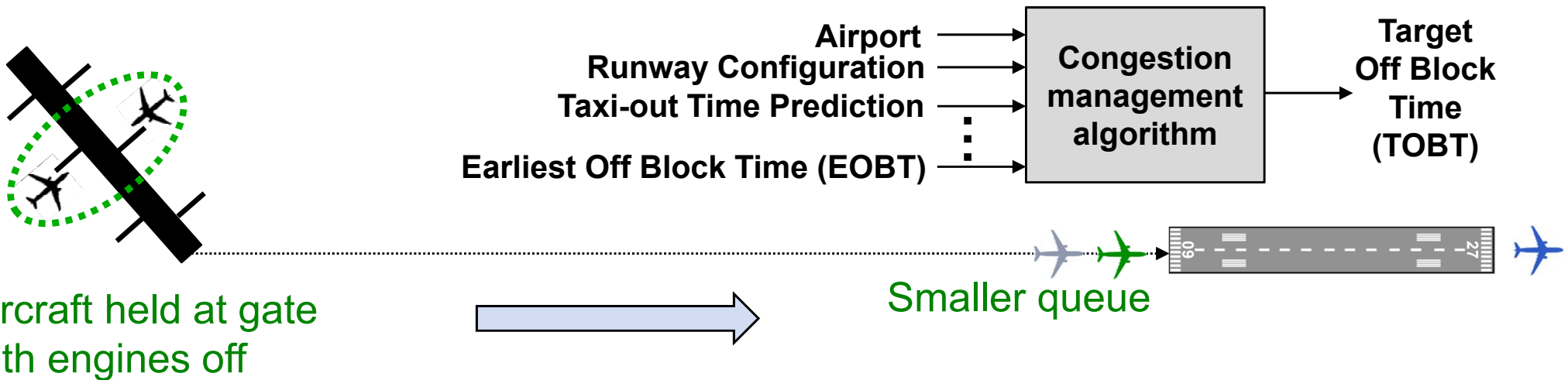
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Default procedures: aircraft push when ready



Congestion management procedure: aircraft publish EOBTs



Aircraft held at gate with engines off

Smaller queue



Outline



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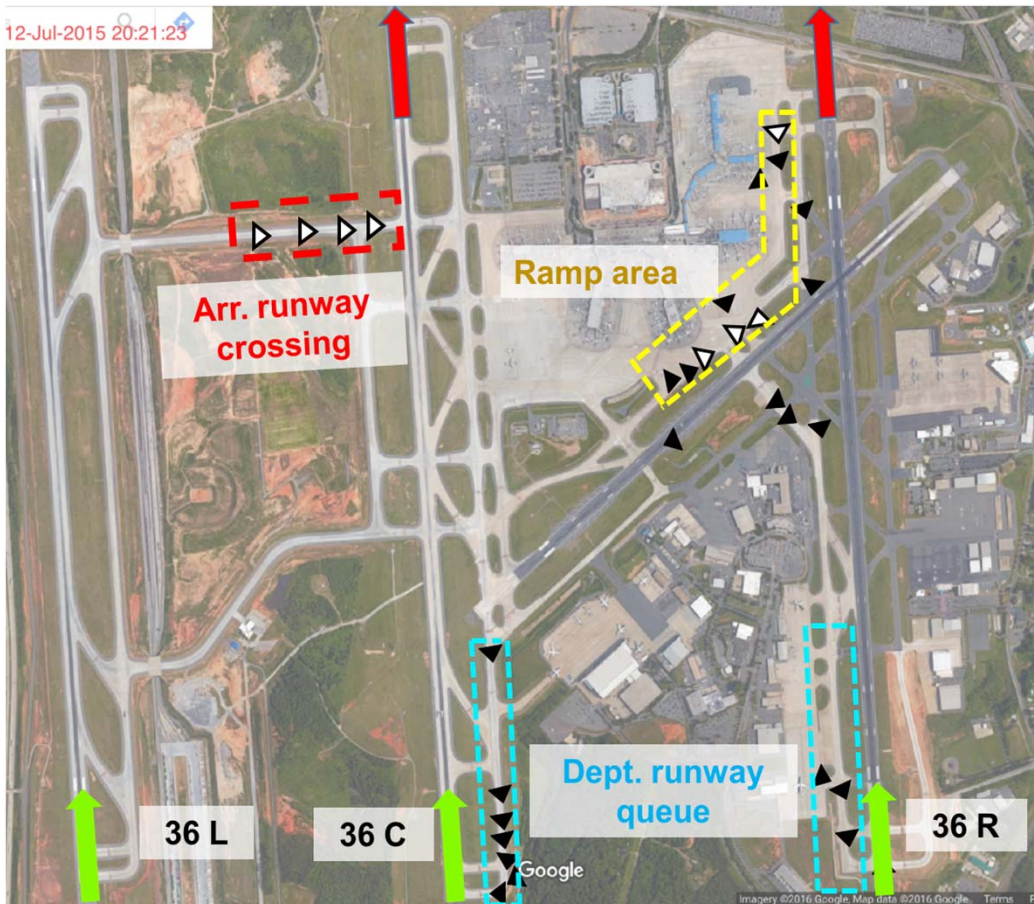
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- Evaluation of a candidate congestion management algorithm using queuing model (NASA's ATD-2) at CLT**
- Estimation of levels of EOBT uncertainty in currently reported data at Charlotte (CLT), Dallas (DFW) & Newark (EWR)**
- Assessment of the impact of EOBT uncertainty on the performance of congestion management algorithm**



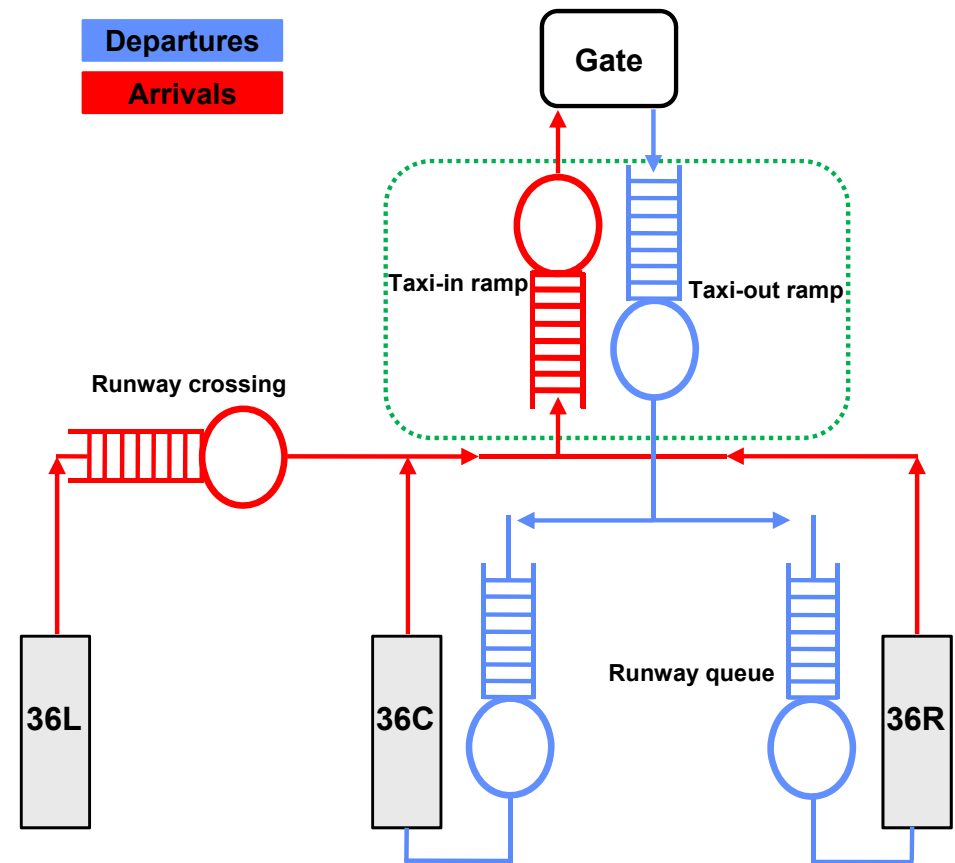
Queuing Network Model of CLT



- Need queuing model to allow with/without surface congestion management comparisons



Charlotte airport layout



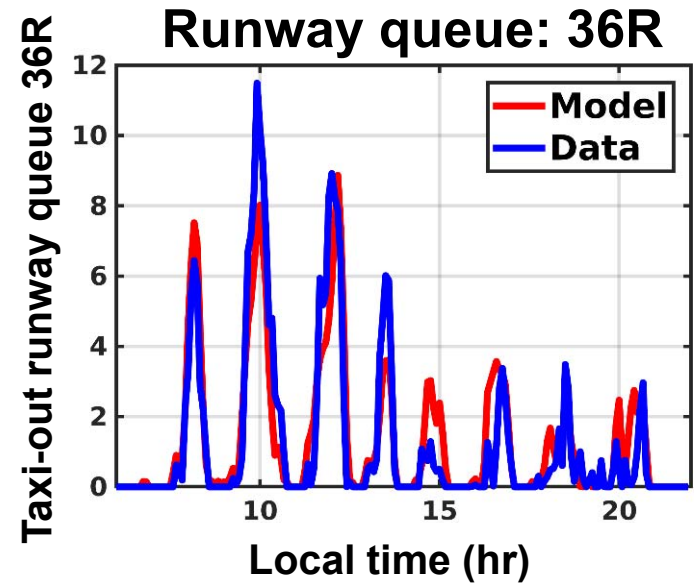
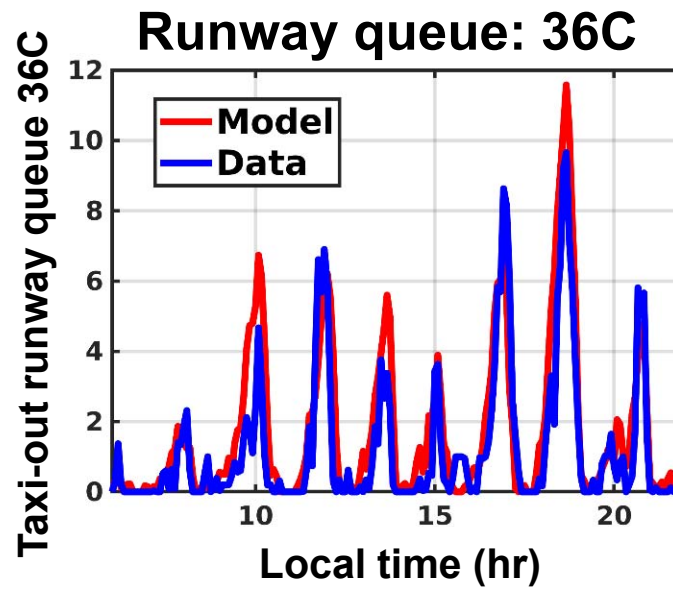
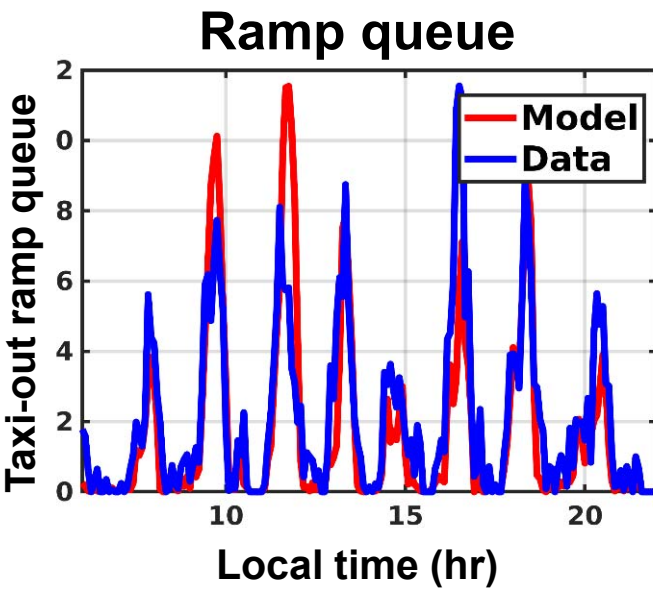
Queuing network representation



Model Validation: Typical Day



- Comparison between the model and data for a typical day (06/25/16)

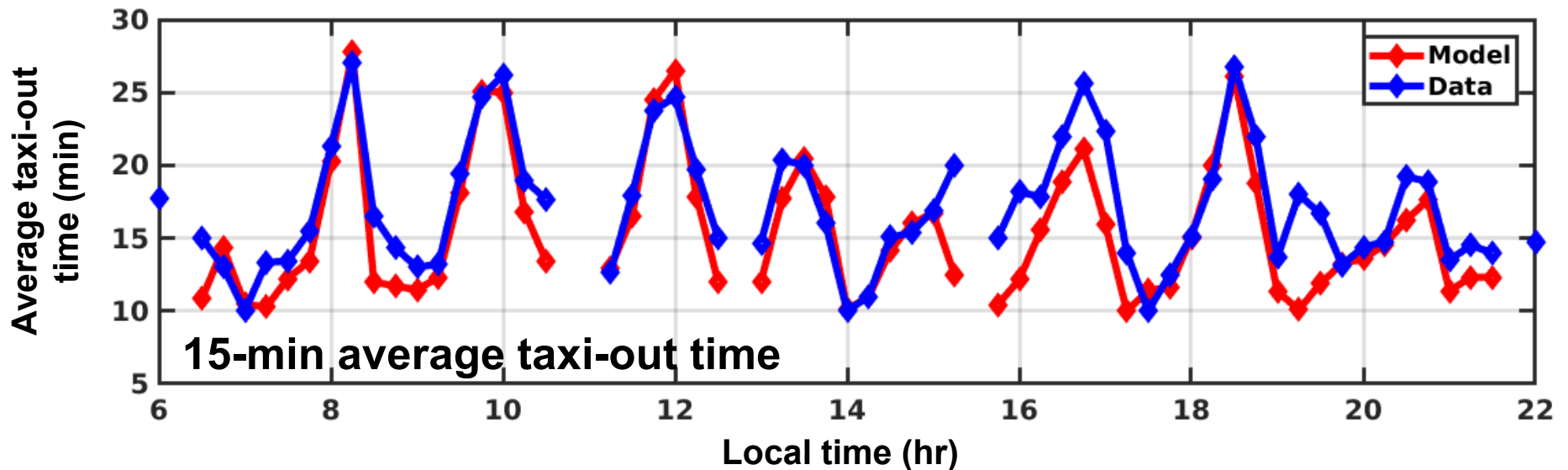




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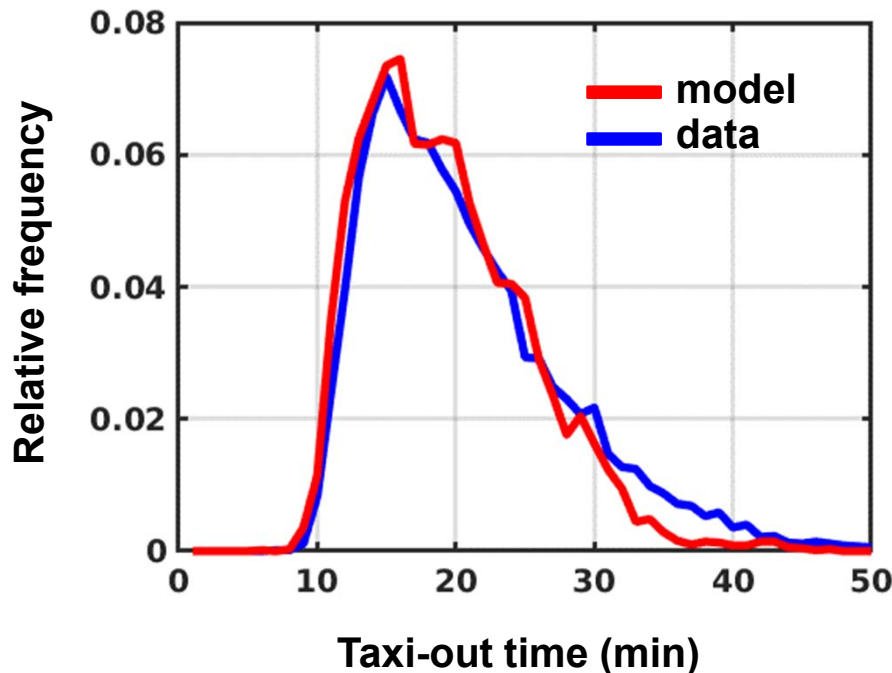




Model Validation: Aggregate Statistics



- Error statistics on an independent test set: 7,484 departures, May/June 2016, CLT northflow



Statistics (min)	Gate to spot	Spot to runway	Taxi-out
Mean values	9.68	10.40	20.09
Mean error	-0.87	-0.58	-1.45
Mean error	3.08	2.7	4.35
% flights with error < 5 mins	82%	86%	69%

- Queuing network model can be adapted to other airports
 - Extended to DFW and EWR for this phase of the analysis



Outline

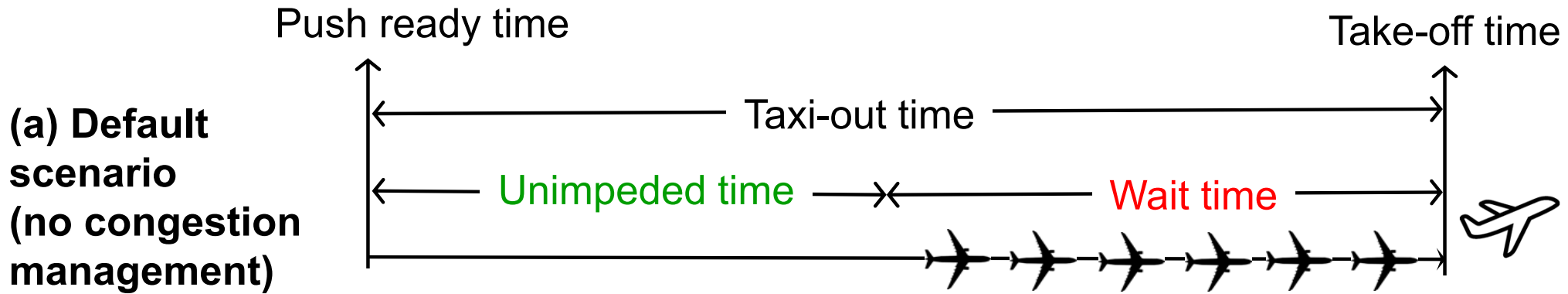


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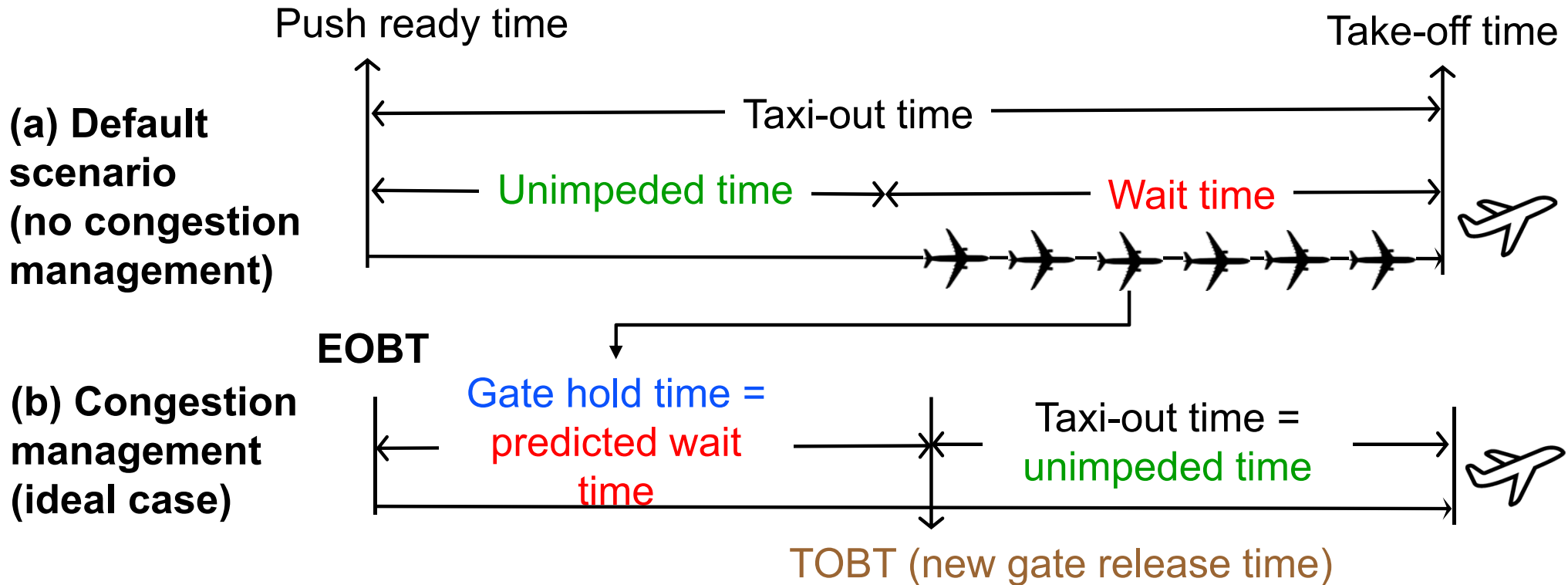


Congestion Management Algorithm: Ideal Case



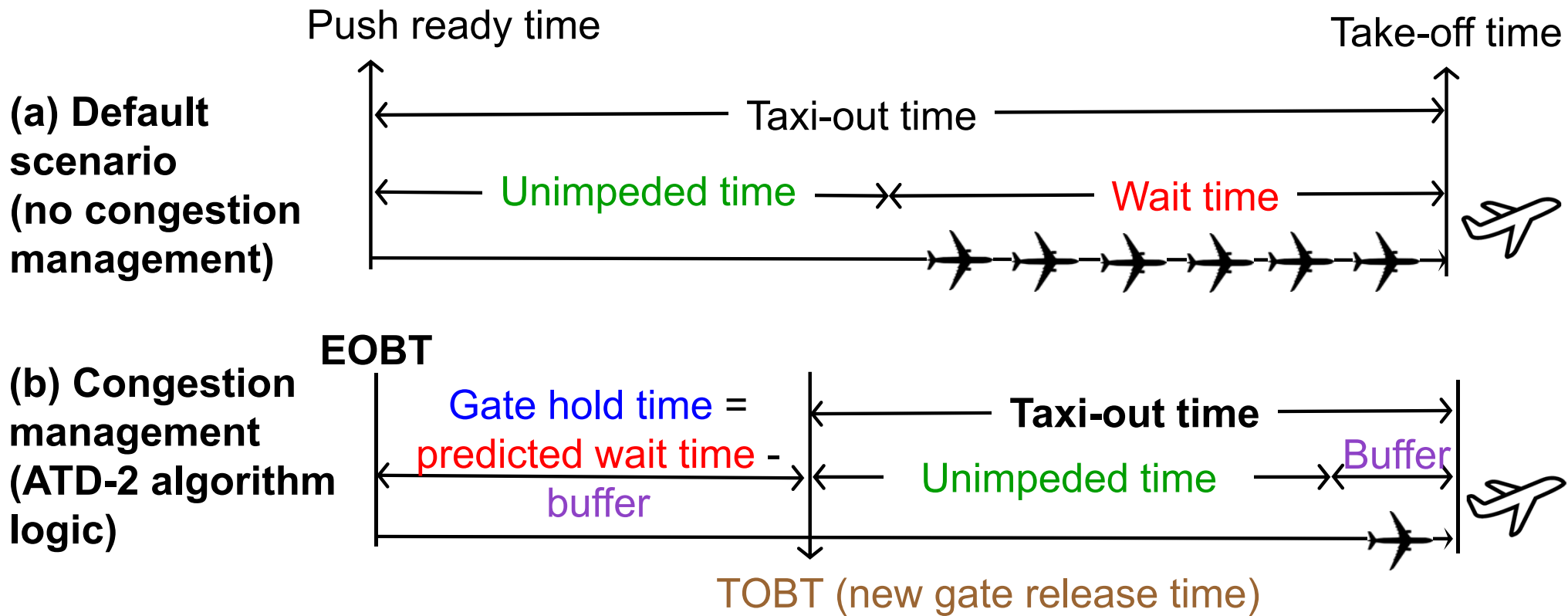


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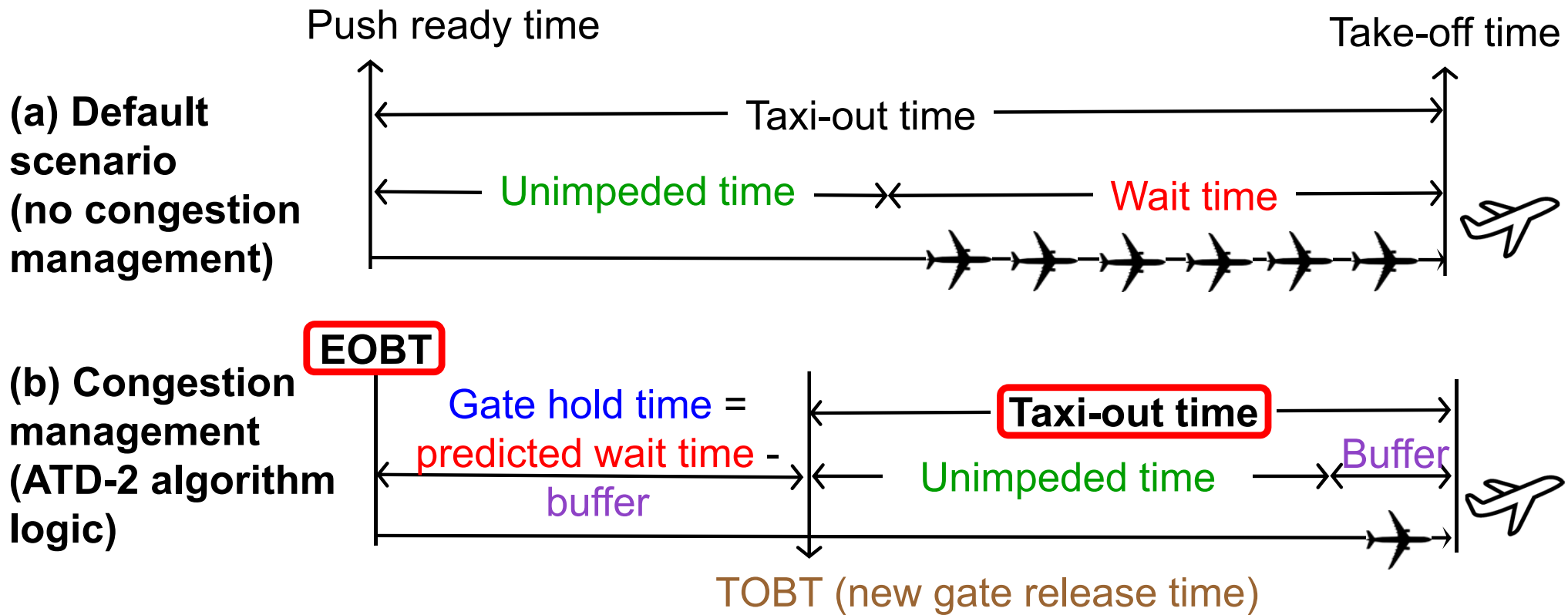
Congestion Management Algorithm: ATD-2 logic



- Buffer parameter accounts for errors in taxi-out time prediction, EOBT and other sources, in order to avoid losing runway utilization
- ATD-2 logic: $TOBT = EOBT + \max(0, \text{Predicted wait time} - \text{Buffer})$



Congestion Management Algorithm: ATD-2 logic



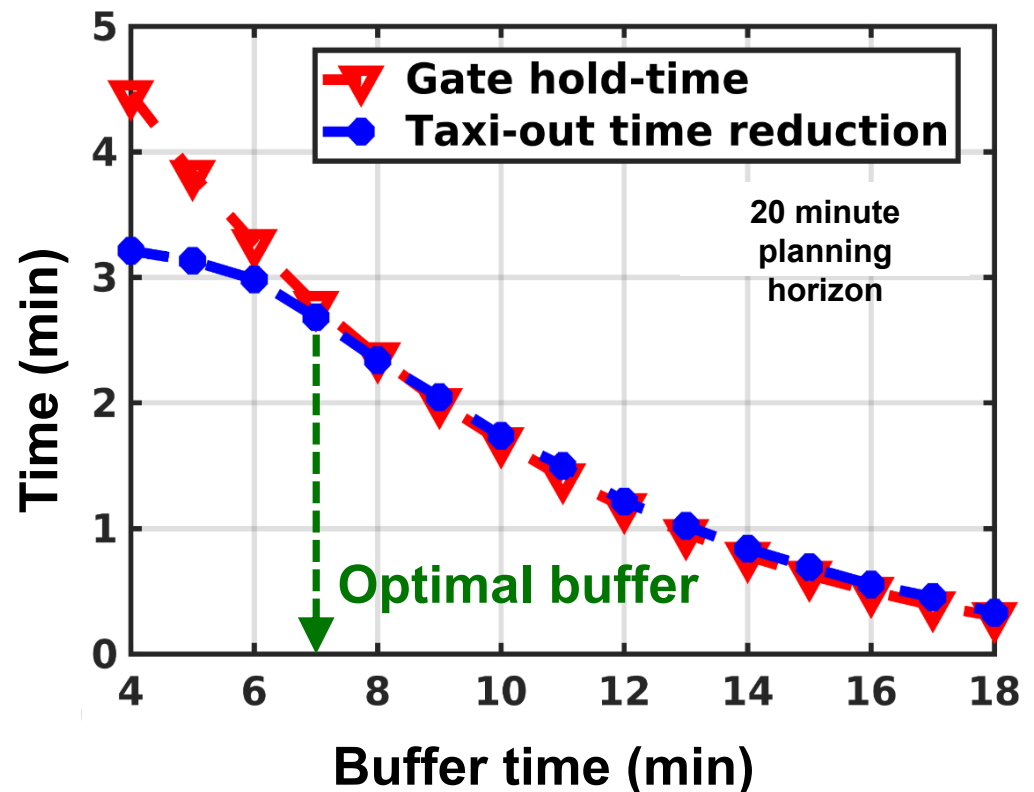
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Congestion Management: Perfect EOBT information



- Departure metering logic tested using stochastic simulations
 - 6,447 departures over 15 day period at CLT
- Taxi-out time reduction depends on the choice of excess queue buffer (larger the buffer, lower the benefits)
- Optimal buffer is lowest value that ensures no loss in runway utilization
- Results with a planning horizon of 20 min





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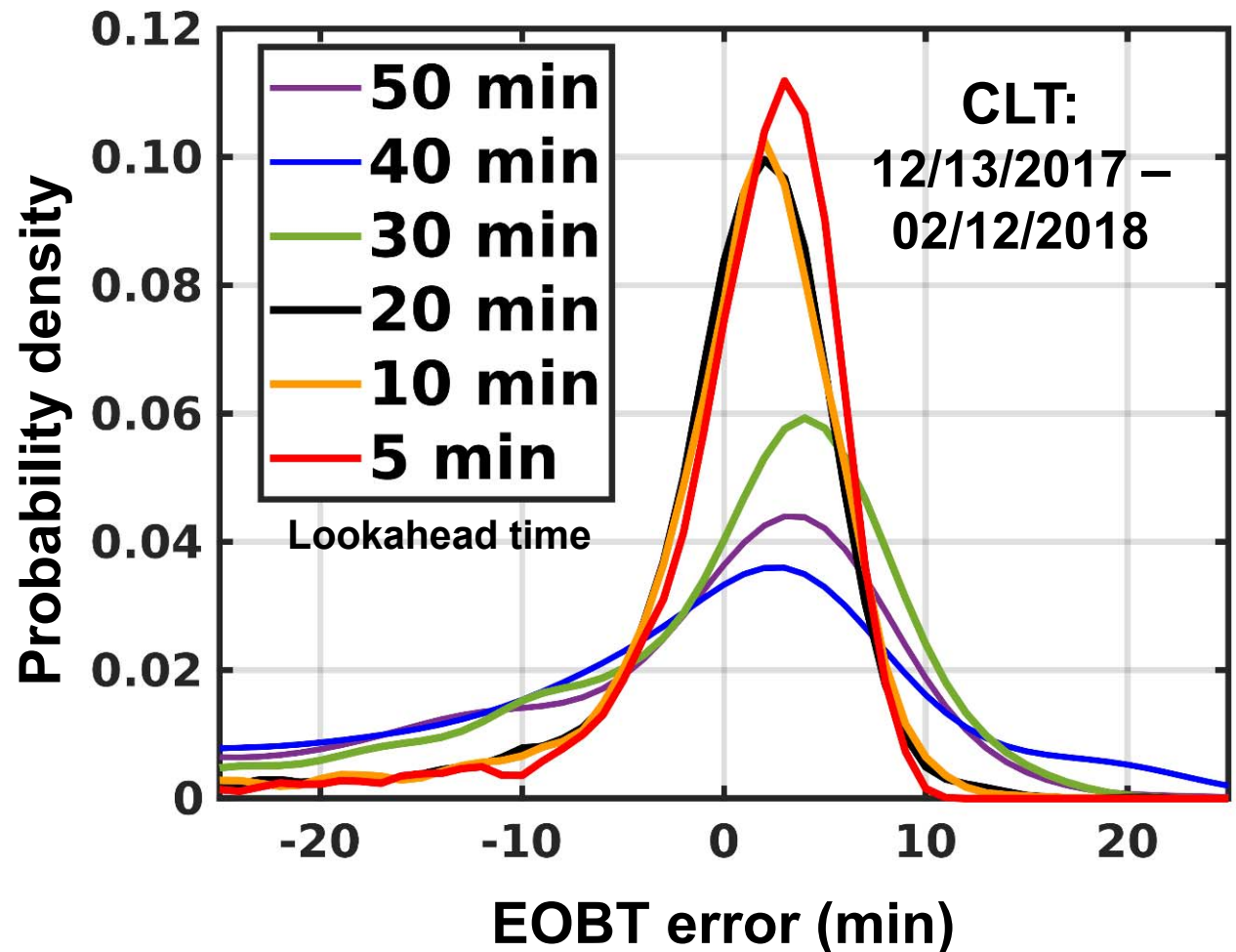
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Evaluation of Empirical EOBT Uncertainty



- Many airlines publish EOBT data through TFMS feed
 - EOBT messages from a major airline shown here
- EOBT error(t) = EOBT(t) – AOBT
- EOBT error varies for different look-ahead times

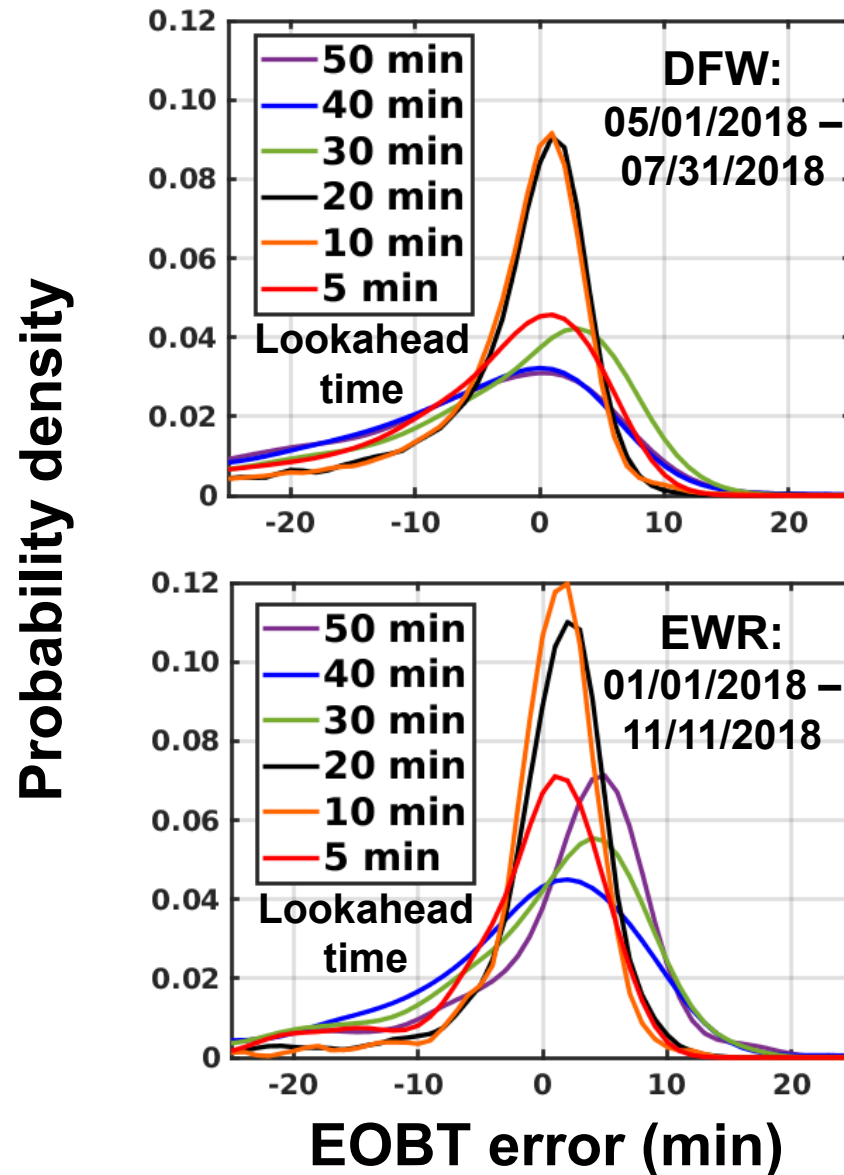




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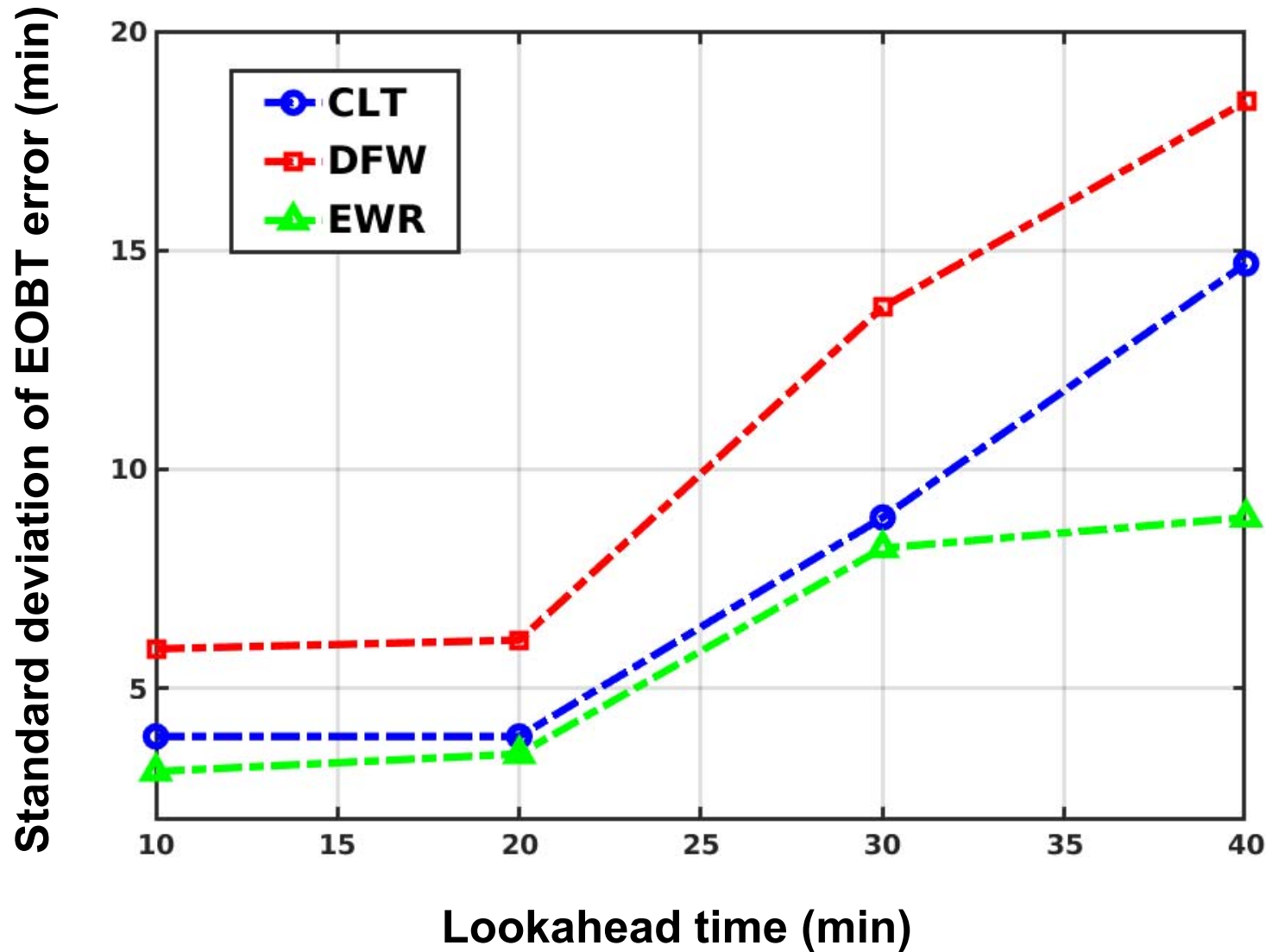


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Evaluation of EOBT Uncertainty Summary Results



- CLT: 12/13/2017 – 02/12/2018
- DFW: 05/01/2018 – 07/31/2018
- EWR: 01/01/2018 – 11/11/2018



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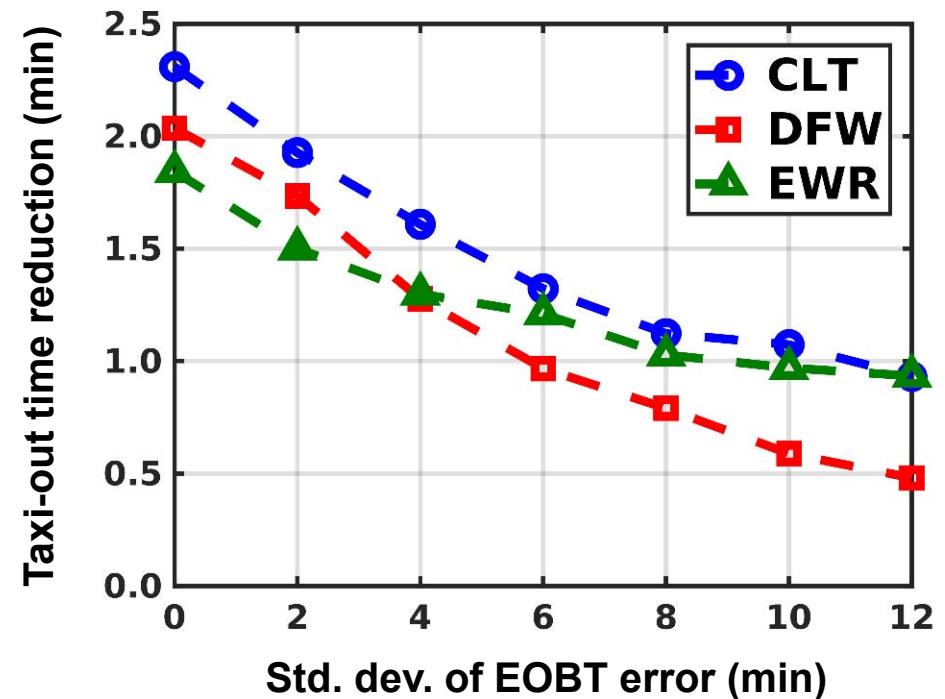
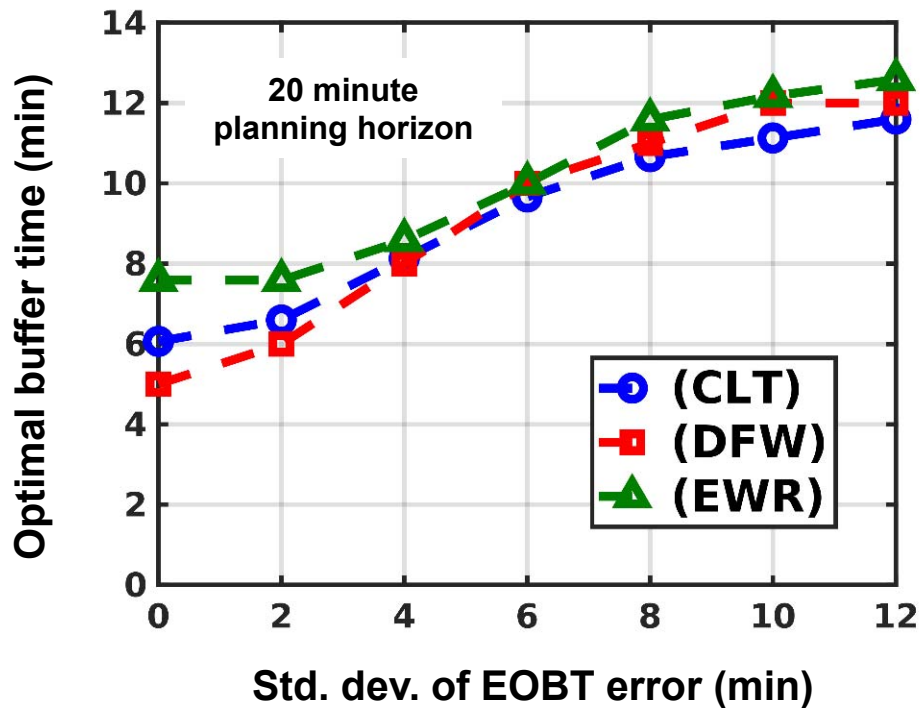
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Congestion Management in the Presence of EOBT Uncertainty

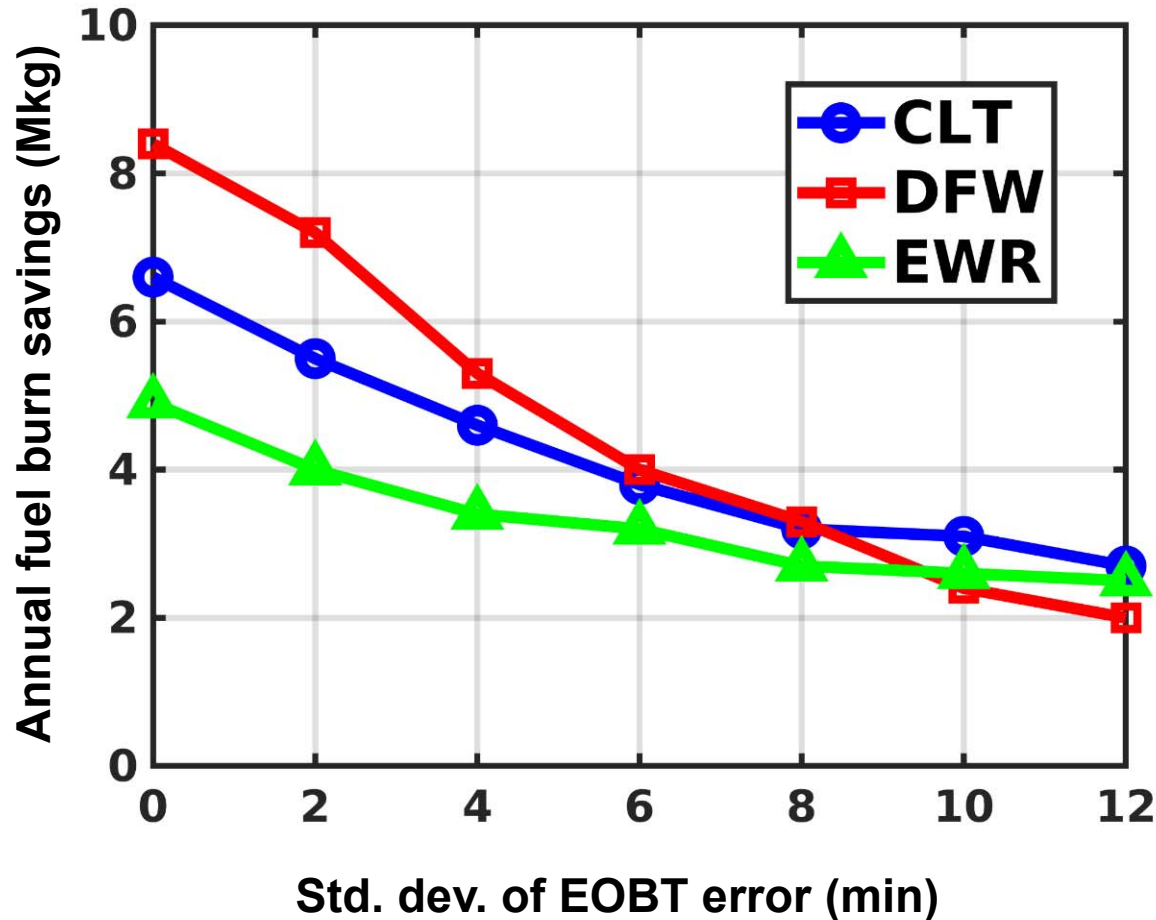


- EOBT uncertainty impacts congestion management because of
 - Reduced prediction accuracy of taxi-out times
 - Non-conformance to the target pushback time
- Need to increase excess queue buffer parameter to account for uncertainties and to maintain runway utilization





Congestion Management in the Presence of EOBT Uncertainty



Taxi-out time & fuel benefits decrease as EOBT error increases



Summary



- **Surface congestion management automation systems will enable fuel and emissions reductions**
- **Analysis presented to understand relationship between automation system benefits and input data (esp. EOBT) accuracy**
- **Informs future algorithm design and airline investment decisions**
- **Recommended next steps**
 - **Extend analysis to broader range of airports and operating conditions**
 - **Analyze incentives for airlines to improve the accuracy of EOBT data**
 - **Develop and evaluate surface congestion algorithms that can**
 - **Explicitly handle uncertainties**
 - **Account for uncertainties in arrival times, in addition to EOBT uncertainties**



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