#### FireDeX: a Prioritized IoT Data Exchange Middleware for Emergency Response

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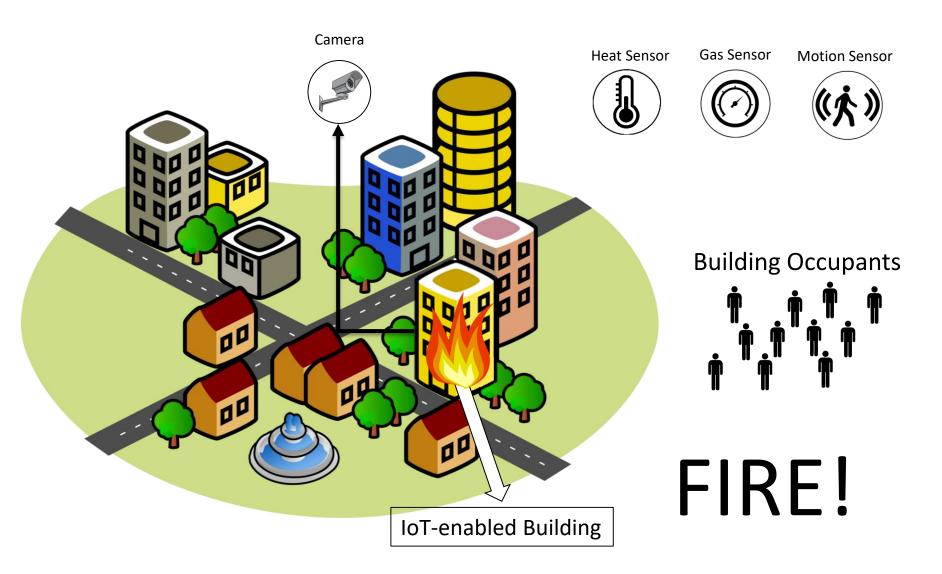
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Joint work with Kyle Benson<sup>1</sup>, Casey Grant<sup>3</sup>, Valérie Issarny<sup>2</sup>, Sharad Mehrotra<sup>1</sup>, Ioannis Moscholios<sup>4</sup>, Nalini Venkatasubramanian<sup>1</sup>

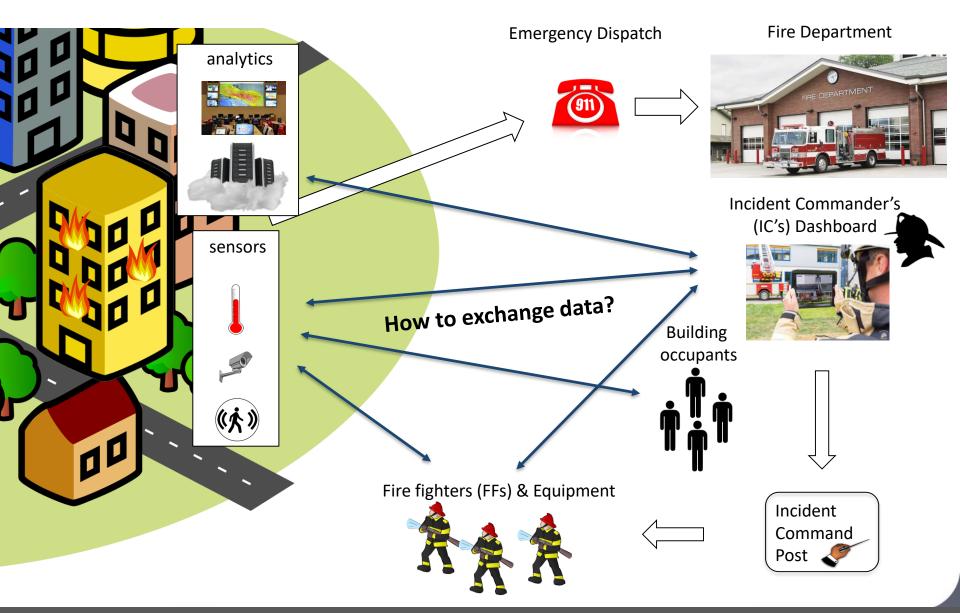
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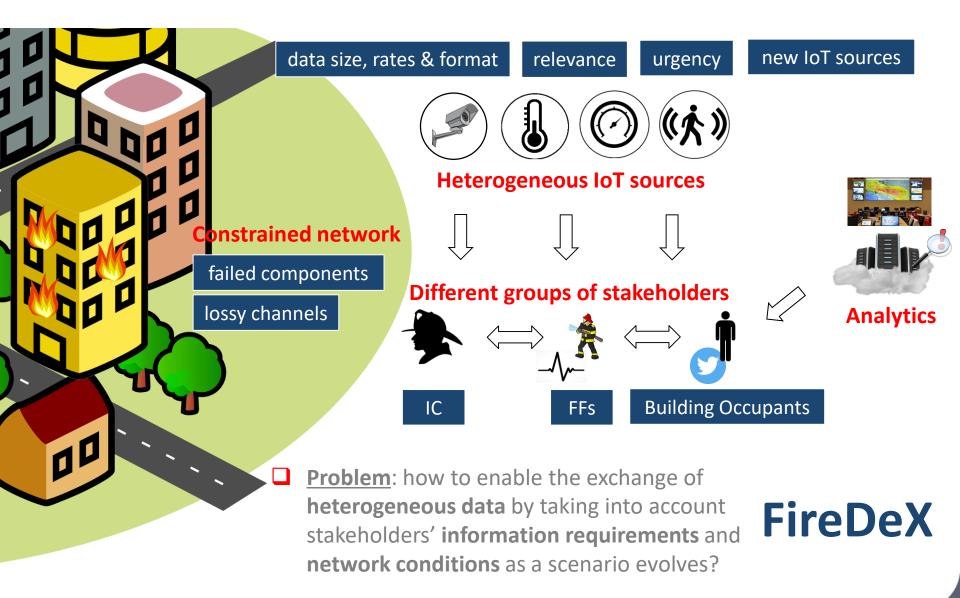
### **Motivation: IoT-enhanced structural fire response**

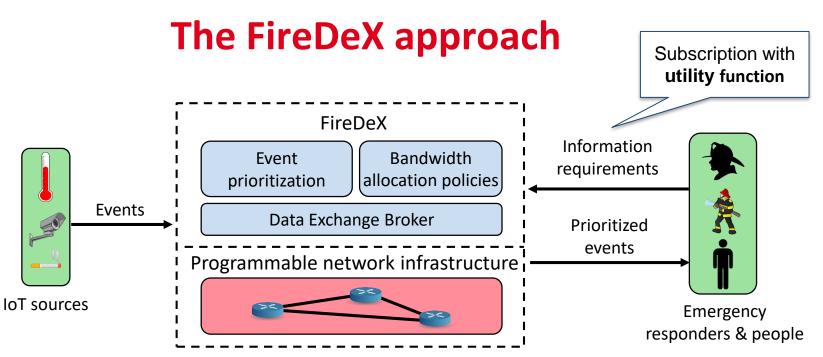


### **Motivation: IoT-enhanced structural fire response**



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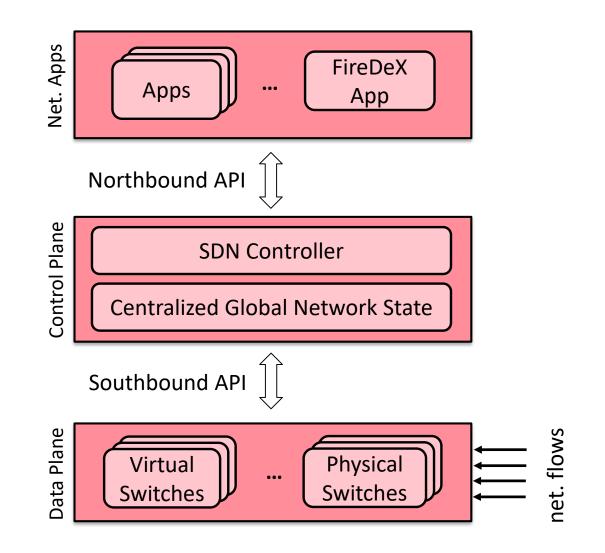


FireDeX middleware configures the data exchange & network with prioritization and bandwidth allocation policies based on:

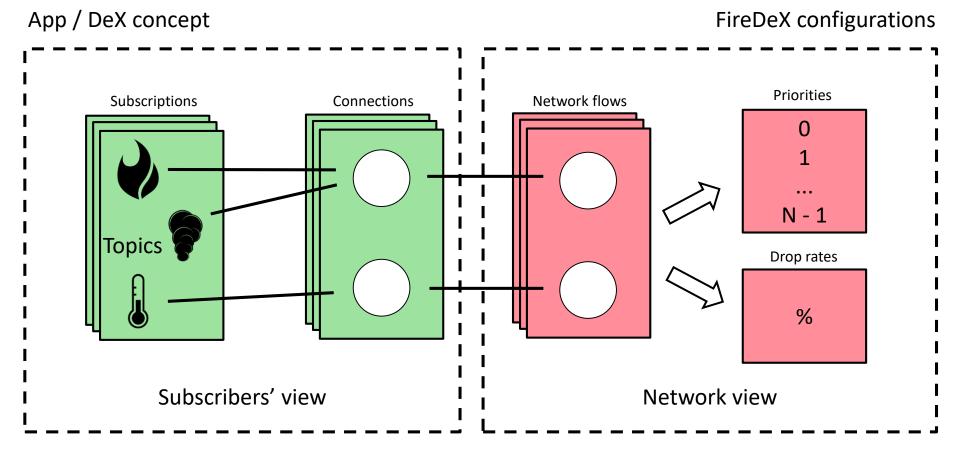
- information requirements
- network resource constraints
- Rely on SDN to bridge critical information requirements with network flows.
- Model the performance of FireDeX across multiple layers using Queueing Theory.
- Use the underpinning formal model for deriving novel algorithms that prioritize IoT events and tune notification delivery/response times.

Goal: timely and reliable delivery of the most critical data to relevant subscribers despite challenging network conditions.

#### **SDN-background**

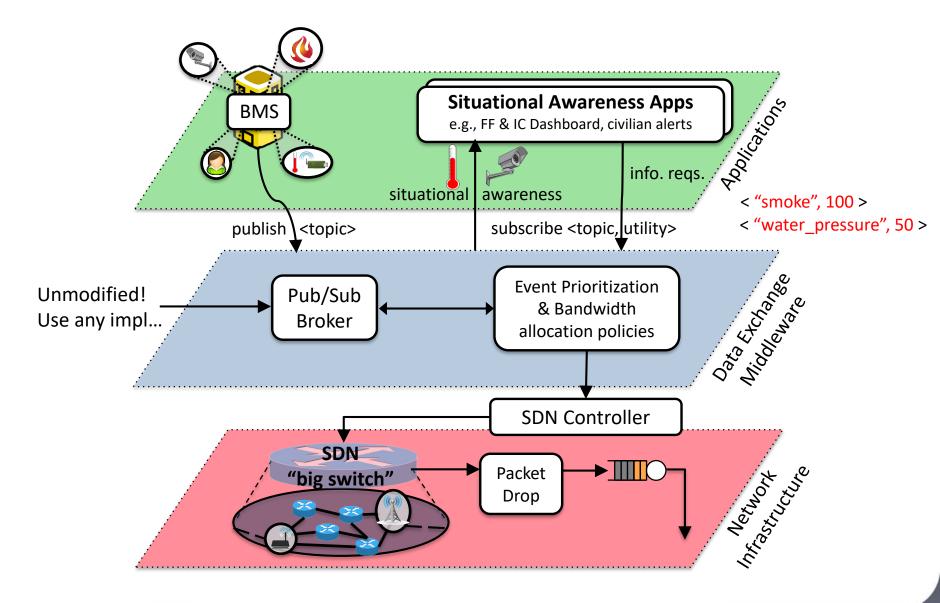


#### Mapping info. reqs. to network state

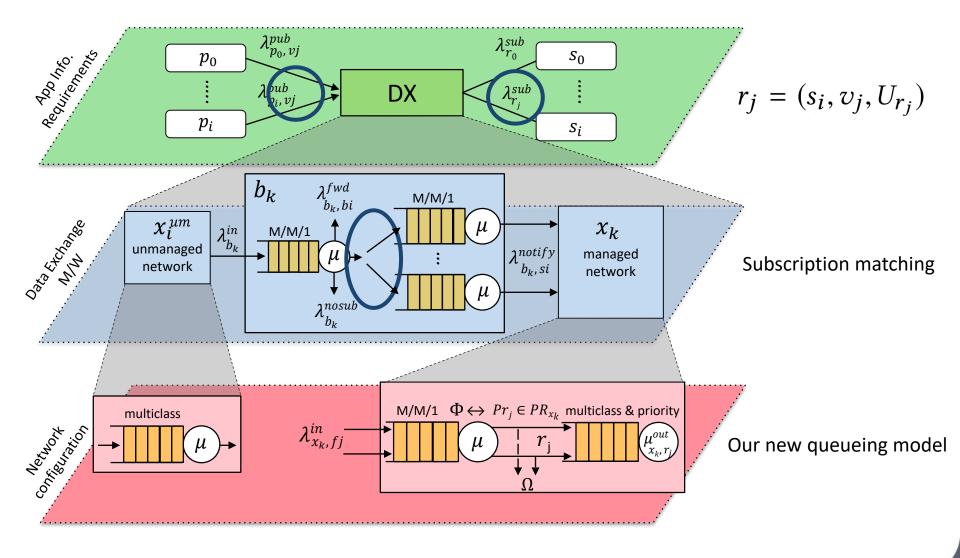


Network flows enable SDN infrastructure to differentiate subscriptions (e.g. by UDP/TCP port number + IP addr).

#### **FireDeX across layers**

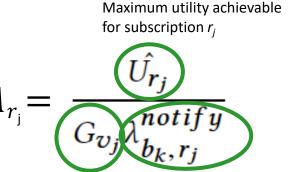


#### **Modeling FireDeX using queuing theory**



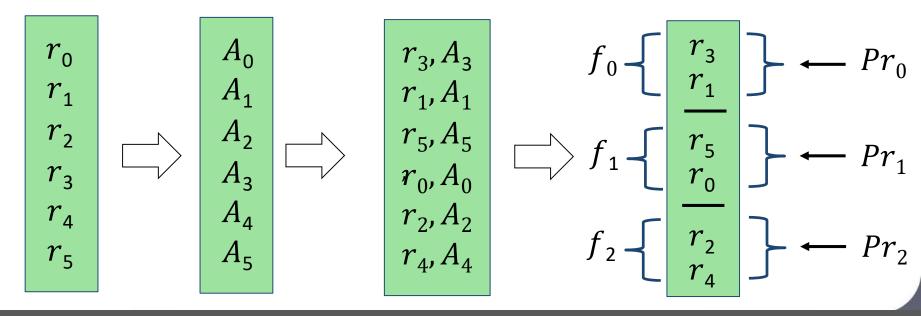
## **Prioritization algorithm**

- Estimate the adjusted utility function per subscription: information value per unit of bandwidth.
- **2.** Sort subscriptions.
- **3.** Group them into approximately equal-sized network flows.
- 4. Priorities assigned to approximately equalsized groups of network flows.



Serialized packet size for topic  $v_j$  notifications

Rate of notifications (publications) for subscription  $r_j$ 



#### **Drop rate algorithms**

Flat

Linear

Exponential

#### Optimized

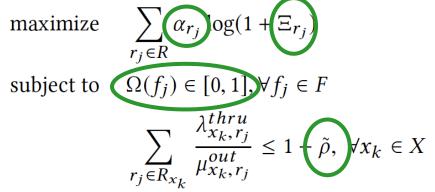
- 1. Formulated as a convex optimization problem.
  - Maximize overall utility as sum of all subscriptions' utilities.
  - Enabled by choice of logarithm for utility function.
- 2. 2nd constraint: queue stability condition.
  - Ensures allocated bandwidth within that available.

Drop rates for each network flow  $\widehat{\Omega}(f_j) = \beta$ 

$$\Omega(f_j) = \beta \Phi(f_j)$$

Mapped priority to network flow

$$\Omega(f_j) = 1 - \beta^{-\Phi(f_j)}$$



"Rho tolerance" enables keeping a buffer within the bandwidth (~0.1)

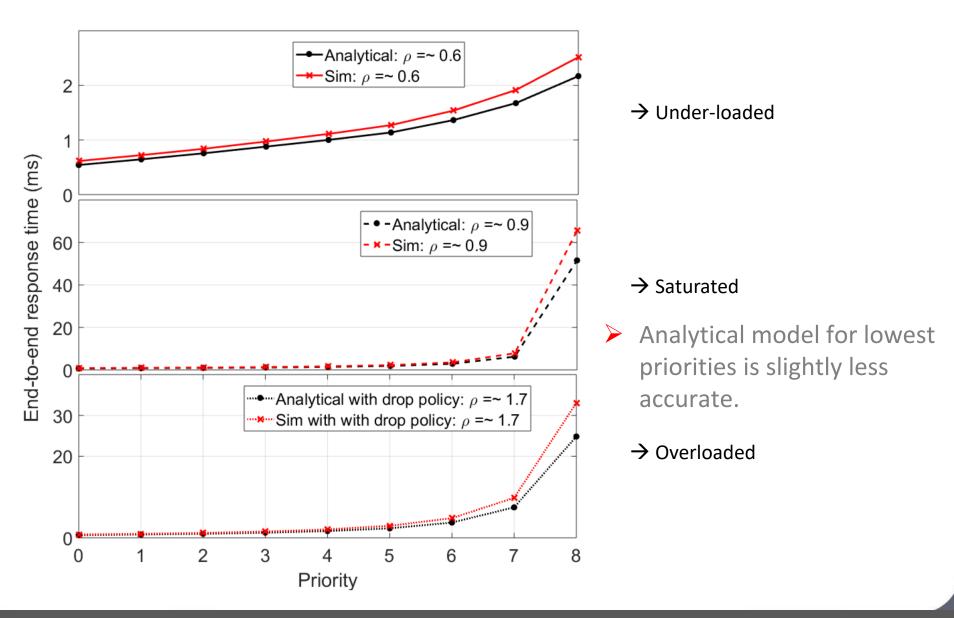
#### **Experimental setup**

- We validate our theoretical model, evaluate the FireDeX approach and compare different prioritization and dropping algorithms.
- We use JINQS (Java Implementation of a Network-of-Queues Simulation) to build our queueing network: an open source simulator for building queueing networks.
  - We have extended JINQS to implement our new multiclass priority queueing model.

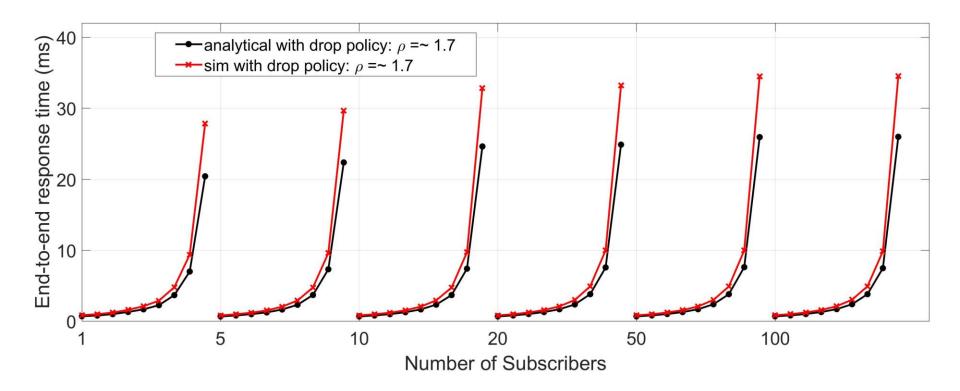
	Data Exchange Parameters								
	#topics ( V )	pub rate $(\lambda_{p_1,p_2}^{pub})$	event size $(G_{v_j})$	#subscriptions $( R_{s_i} )$	<b>utility</b> <b>weight</b> $(\alpha_{r_i})$				
Telemetry data	140	$Exp(\frac{1}{6}) \in [4,7]$	$Exp(\frac{1}{110}) \in [90,500]$	70	$   \operatorname{Exp}(\frac{1}{0.5}) \in \\     [0.01,2] $				
Async. events	60	$Exp(\frac{1}{4}) \in [3,5]$	$Exp(\frac{1}{800}) \in [500, 1100]$	42	$Exp(1) \in \\ [0.1,4]$				

Network Parameters									
#subscribers	<b>#publishers</b>	#flows	<b>#priorities</b>	bandwidth	$\rho$ tolerance				
(  <i>S</i>  ) 10	(  <i>P</i>  ) 160	$( F_{s_i} )$	9	$(w_{s_i})$ 80 Mbps	( <i>ρ</i> ) 0.1				

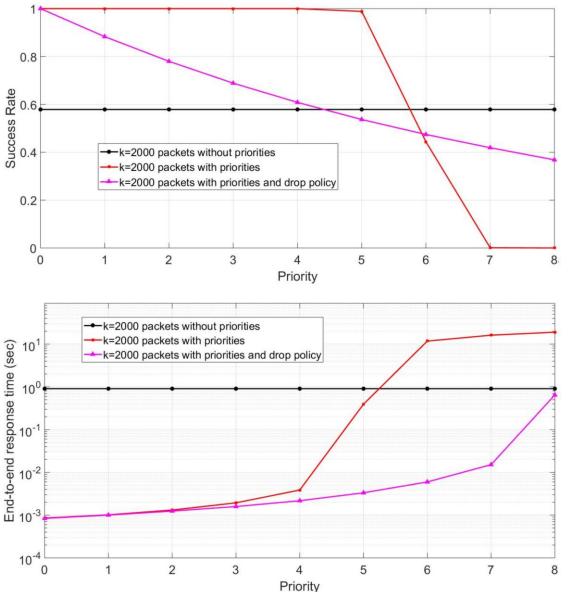
### Model validation: varying traffic loads



#### **Model validation: scalability**



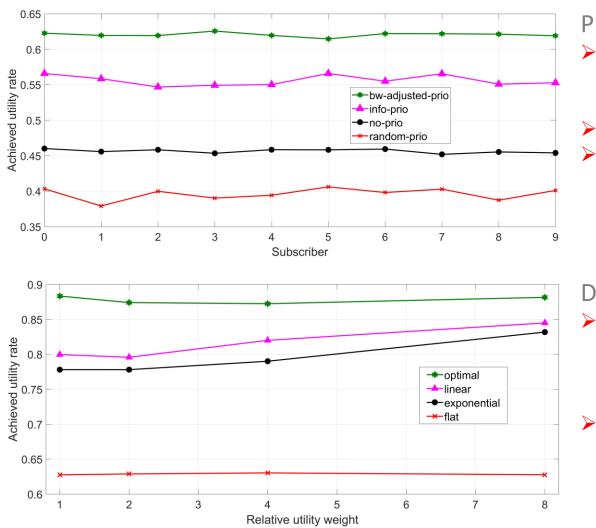
#### **FireDeX approach evaluation**



- With an overloaded system, switch buffers fill up and cause high delay / packet drops.
- Our approach delivers more high priority events than finite buffers only.

- High priority events also delivered quicker.
- Addition of drop rate policy smooths success rate while reducing end-to-end delay.

## **Algorithms comparison**



Prioritization algorithms:

- Our bandwidth-aware greedy strategy performs better than bandwidth-unaware version.
- Both better than **no prioritization**.
  - But **random** priorities are worst: need to set priorities correctly!

#### Drop rate algorithms

- Convex optimization performs best in comparison to linear, exponential and flat policies (drop rates by assigned priority).
  - Plot shows **varying utilities** of async events vs. data telemetry: simpler policies perform closer to optimal for larger differences.

#### **Conclusions & Next steps**

#### Conclusions

- We introduce a middleware that integrates application and network awareness.
- Our application-aware prioritization algorithm improves the value of exchanged information by 36% when compared with no prioritization.
- Network-aware drop rate policies improve this performance by 42% over priorities only and by 94% over no prioritization.

#### **Next steps**

Queueing model:

 Consider non-Poisson arrival and service rates by using G/G/1 or G/D/1 queues.

System:

- Alternative utility functions.
- Tuning the entire broker network.
- Use our TIPPERS testbed and CFAST simulator to further evaluate the FireDeX approach.

# Thank you







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