



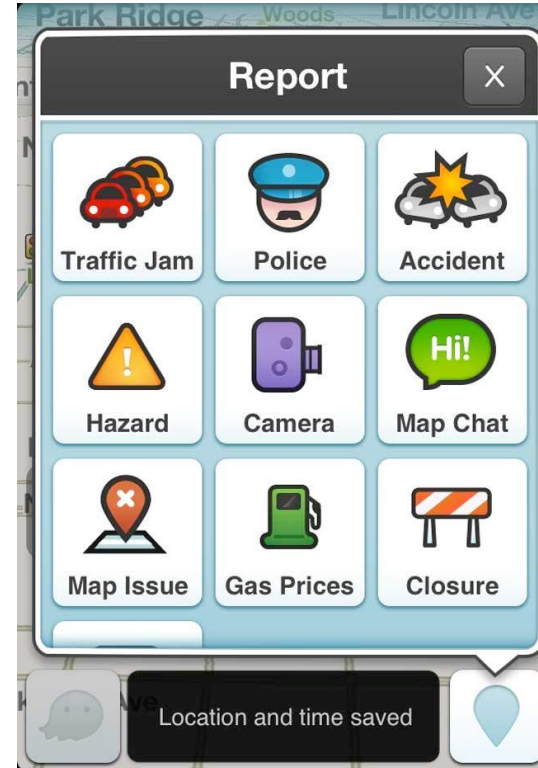
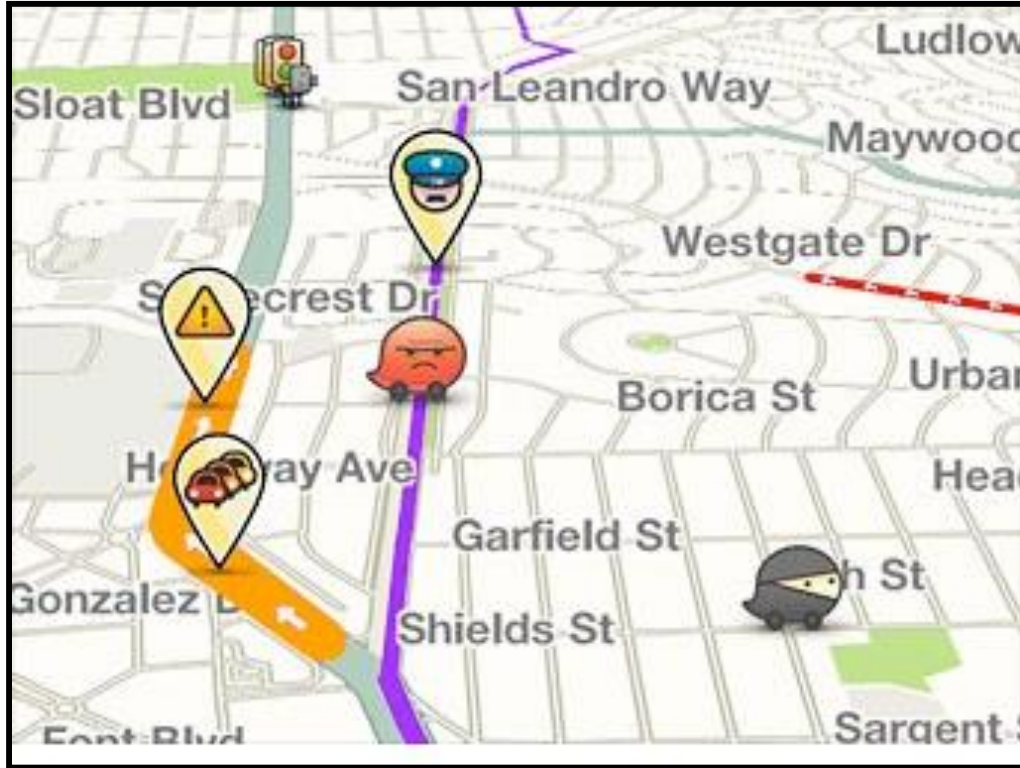
# Detecting Repurposing and Over-collection in Multi-Party Privacy Requirements Specifications

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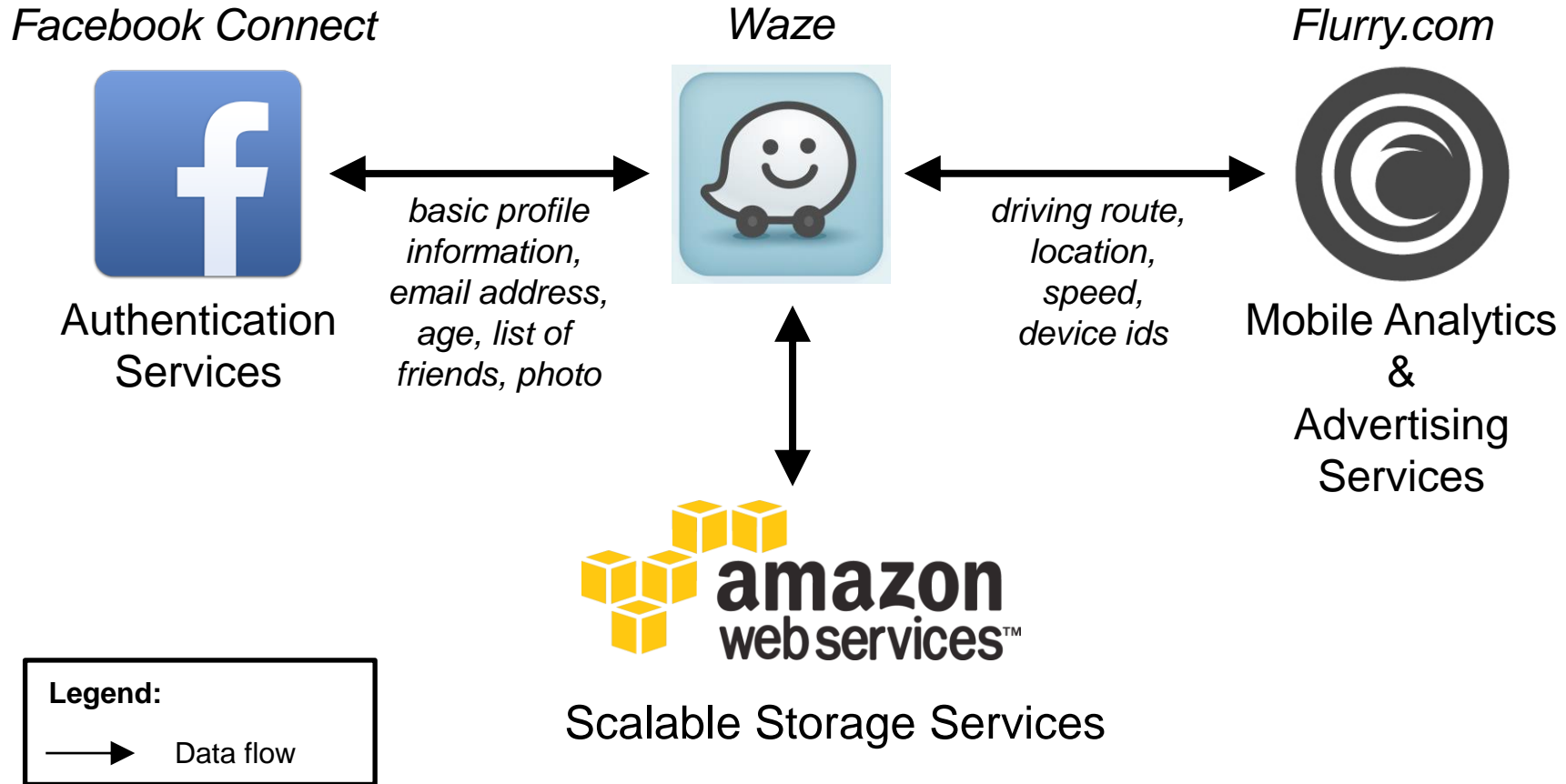
# Motivation

- Mobile and web applications are using Service Oriented Architectures more and more.
- How do we ensure that using 3<sup>rd</sup> party services doesn't increase our privacy risk?
  - What sort of data do they want?
  - What will they do with my data once they have it?
  - What am I willing to give them?
- Tough to answer these questions.

## What is **waze** ?



# Example Service Integration



# Case Study Research Questions

**RQ1:** What conflicts exist in our formalization of Waze's policies?

**RQ2:** What multi-party data flows exist?

**RQ3:** Does data repurposing or over collection occur?

# Building on Previous Work

- [BR13] introduced Eddy.
- SQL-like syntax for policy specifications.
- Limited to tracing policies within a system; can't extend to 3<sup>rd</sup> parties.
- Great for finding conflicts in policies (conflicting interpretations).
- Some performance analysis.

[BR13] T.D. Breaux, A. Rao. "Formal Analysis of Privacy Requirements Specifications for Multi-Tier Applications, *21<sup>st</sup> IEEE International Requirements Engineering Conference*, pp. 14-23, Jul. 2013

# Related Work

- Extracting goals from privacy policies  
[Antón et al.,2004; Breaux & Antón, 2005; Young et al., 2011]
- Formal models of privacy-related requirements  
[Breaux, Hibshi, Rao, 2013; Liu et al. 2003; Tun et al. 2012; Omoronyia et al., 2013]
- Static and dynamic analysis of code (TaintDroid, Appfence, Pscout)  
[Enck et al., 2010; Hornyack et al. 2011; Yee Au et al. 2012]
- Multiple policy-related languages...

# The Value of Knowing

## Maximize Data Utility

- Collect everything, value is realized later
- Ensure open access; this drives innovation
- Disclose to leverage third-party value
- Retain as long as practical (longitudinal/behavioral)
- Avoid destruction





# Balancing Utility and Risk

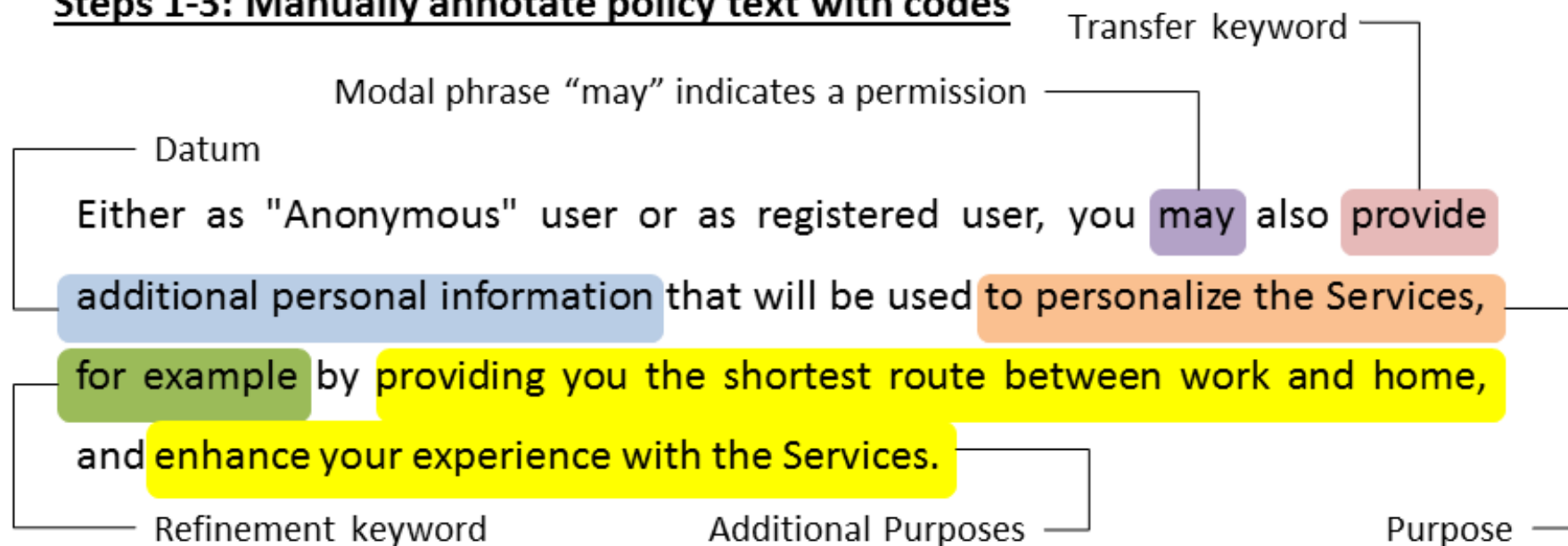
## Maximize Data Utility

- Collect everything, value is realized later
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## Minimize Privacy Risk

- Limit collection based on stated needs
- Limit access, obtain consent for new uses
- Limit disclosure and third-party uses
- Destroy when no longer needed
- Embrace destruction

### Steps 1-3: Manually annotate policy text with codes



### Step 4: Write expression in Eddy (re-topicalized for Waze)

SPEC-HEADER

P **personalizing-services** > **providing-shortest-route, enhancing-service-experience**

SPEC-POLICY

P **COLLECT** **personal-information** FROM **waze-user** FOR **personalizing-services**

### Step 5: Tool compiles Eddy into Description Logic

- (A) **providing-shortest-route**  $\sqsubseteq$  **personalizing-services**
- (B) **enhancing-service-experience**  $\sqsubseteq$  **personalizing-services**
- (C)  $p_6 \sqsubseteq$  **COLLECT**  $\sqcap$   $\exists$ **hasObject.personal-information**  $\sqcap$   
 $\exists$ **hasSource.waze-user**  $\sqcap$   $\exists$ **hasPurpose.personalizing-services**
- (D)  $p_6 \sqsubseteq$  **Permission**

# Three Privacy Principles

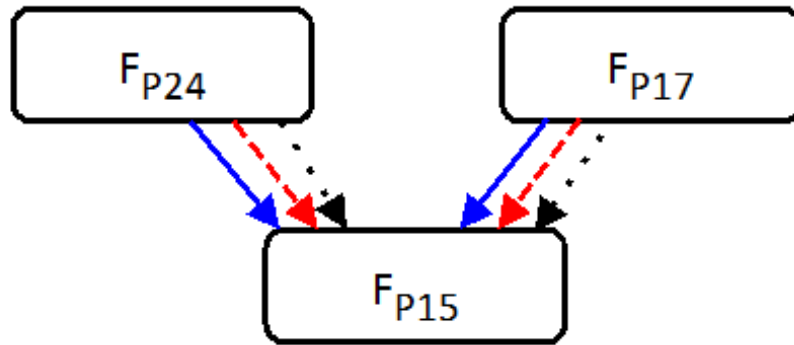
- *Purpose specification principle:*
  - The purposes for which data is collected should be explicitly stated.
- *Collection limitation principle:*
  - Collection of personal data should be limited (to what will be used).
- *Use limitation principle:*
  - Uses should be limited to the purposes for which the data was originally collected, and nothing else.
- Exceptions for consent and legal compliance.

# Three Privacy Principles

- Commonly accepted.
  - U.S. Fair Information Practice Principles (FIPPs)
  - OECD Guidelines on the Protection of Privacy and Trans-border Flows of Personal Data
- If these principles are violated, there are privacy risks.
  - Repurposing
  - Overcollection

# Crossflow Analysis (1<sup>st</sup> Party), Identifying Data Over- and Under-flows

P COLLECT **device-id, ip-address**, P.COLLECT **device-id, location, ...**  
FROM **end-user** FOR anything      FROM **application** FOR anything



P TRANSFER **device-id**  
FROM **anyone** FOR anything

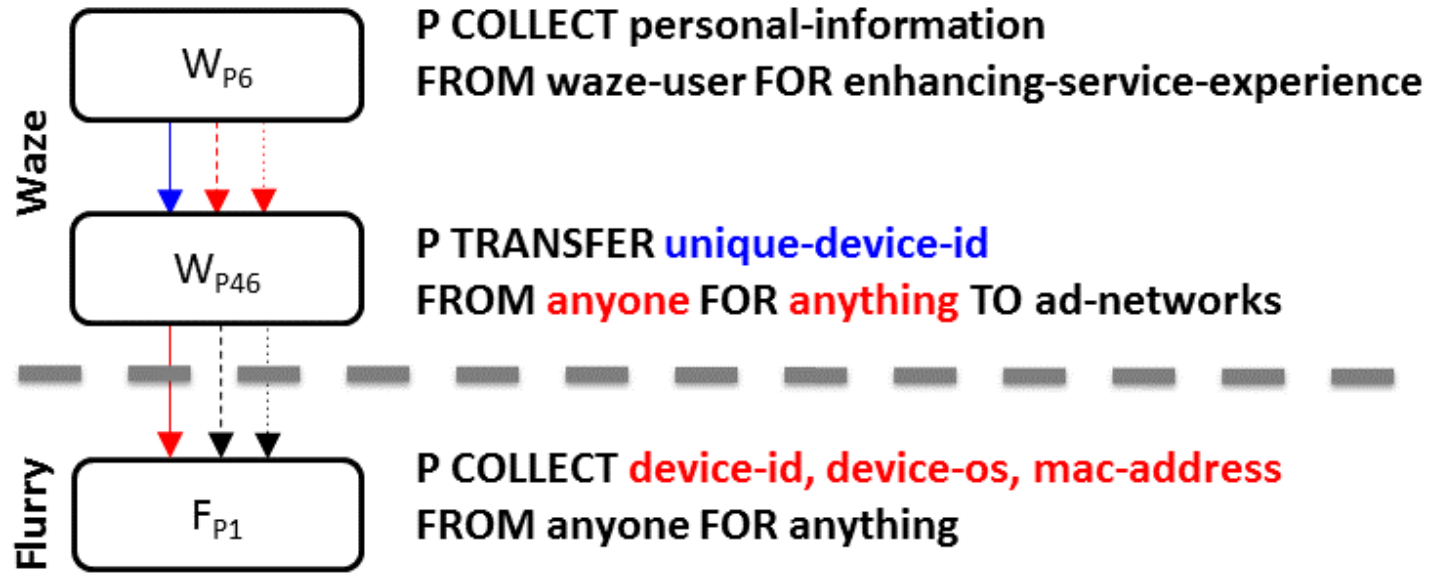
Legend:

- ← hasObject
- ←--- hasSource
- ←..... hasPurpose
- Blue**: overflow
- Red**: underflow
- Black**: exact flow

# Tracing to 3<sup>rd</sup> Parties

- Requires a dictionary, to map each party's lexicon.
- Your definition of information is different to mine.
- Your definition of a purpose is different to mine.
- And so on...
- Dictionaries can be developed separately by different parties.

# Crossflow Analysis (3<sup>rd</sup> Party)



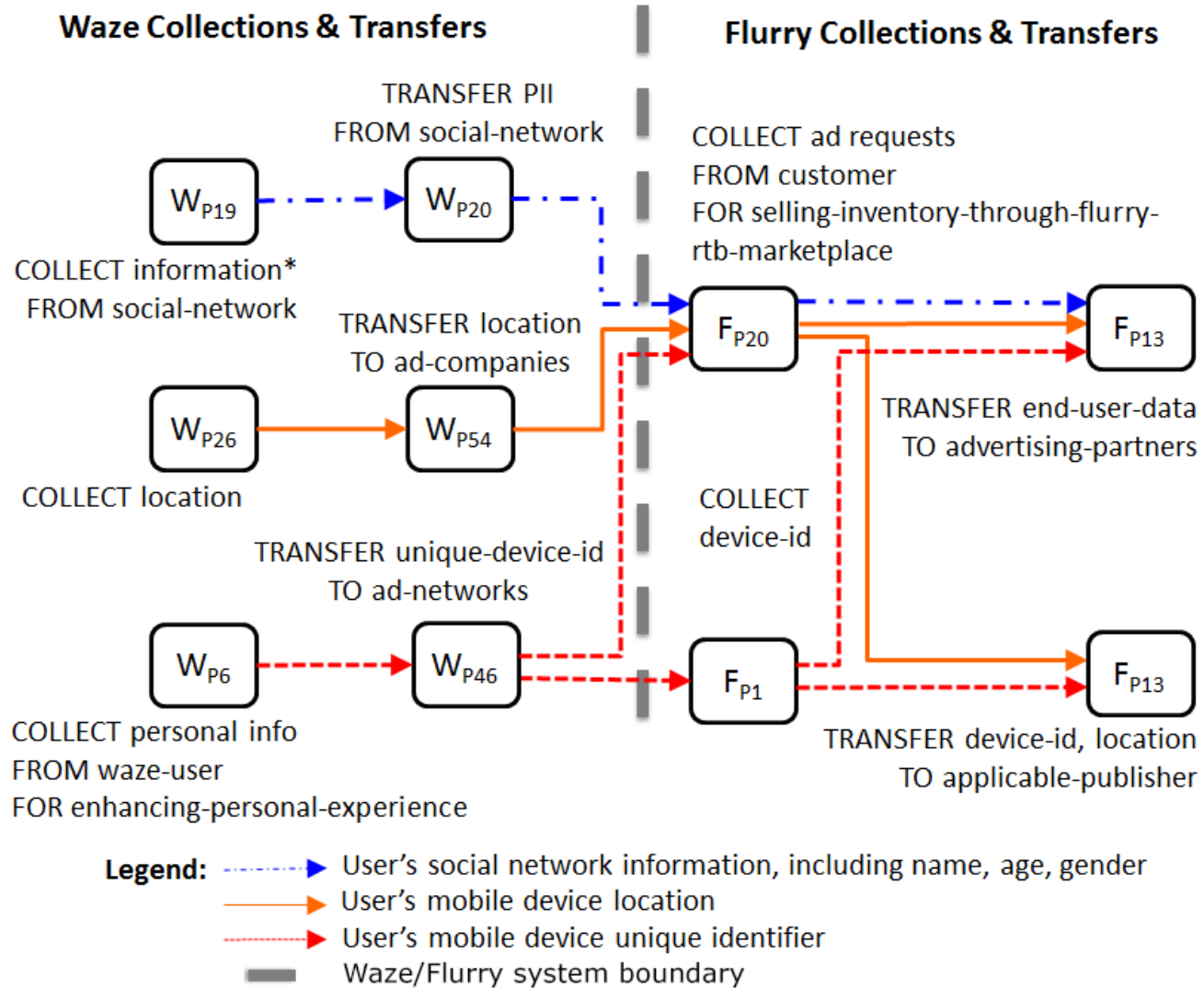
Legend:

- ← hasObject
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- Blue:** overflow
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▬ Waze/Flurry system boundary

## Crossflow Analysis (3<sup>rd</sup> Party)





## Waze Case Study Results

Policy	Total Stmts	Data Req'ts	Modality <sup>1</sup>				Actions <sup>2</sup>	
			P	O	R	C	U	T
Waze	150	65	60	0	5	13	18	34
Flurry	155	44	42	0	2	15	6	23
Facebook	136	55	24	1	30	13	24	18

### Overview of Requirements

1. Privacy policies generally describe permissions (P), with few prohibitions (R) and almost no obligations (O)
2. Data requirements describe only collect (C), use (U) and transfer (T) actions, which comprised 28-43% of total policy

**Patterns: (Purpose Hoisting, Unrestricted Cross-Flows)**

# Waze Case Study Results

Policy	Definitions		Axioms				Concepts	
	Expl.	Impl.	S	D	E	A	D	P
Waze	19	29	41	3	4	6	29	13
Flurry	14	20	21	1	12	0	34	0
Facebook	13	0	11	0	2	0	13	0

## Ontology Complexity

- Inferences to discover implied (Impl.) definitions (e.g., *personal information* is equivalent to *personal details*).
- Formalisms: Subsumption (S), Disjointness (D) and Equivalence axioms (E).
- Concepts: Actors (A), Data types (D) and Purposes (P)

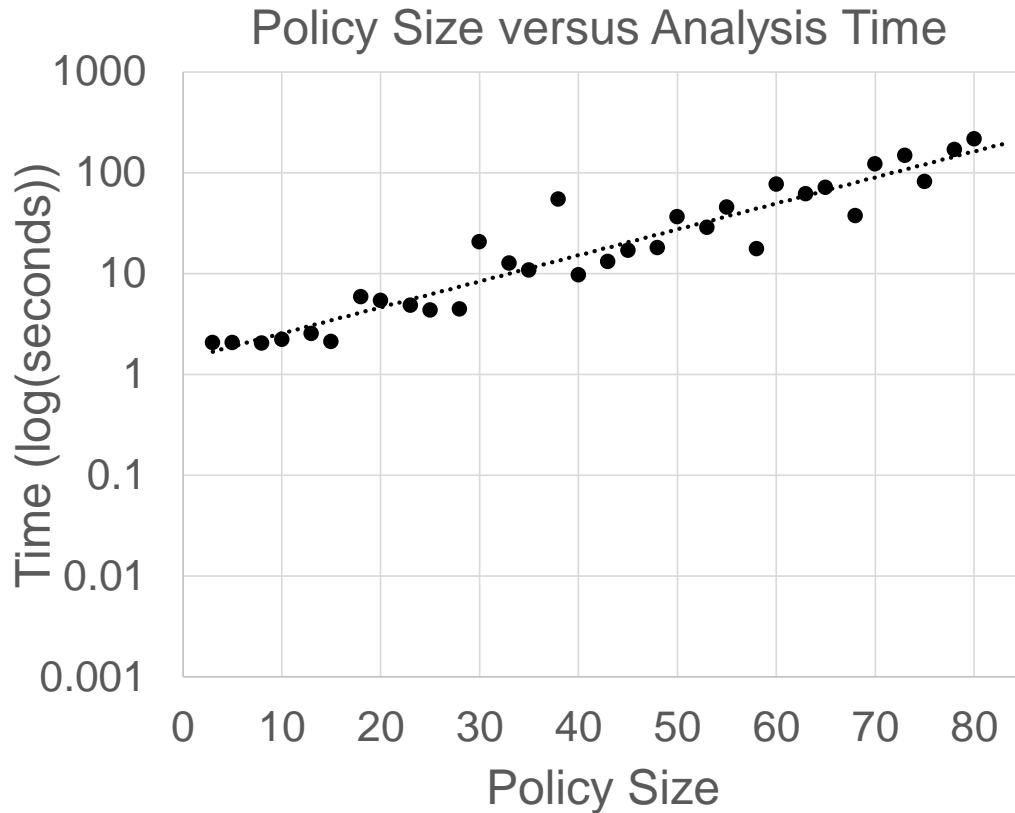
# Waze Case Study Results

- Found conflicts – envision tool-driven techniques to help specification authors to detect and remove these.
- Discovered attacks on collection and use limitation principles (unrestricted crossflows, purpose hoisting)
- Scaling tools to larger policies (akin to map-reduce).
- Expand to other data practices: consent, data retention, etc.

# Quick note on performance analysis...

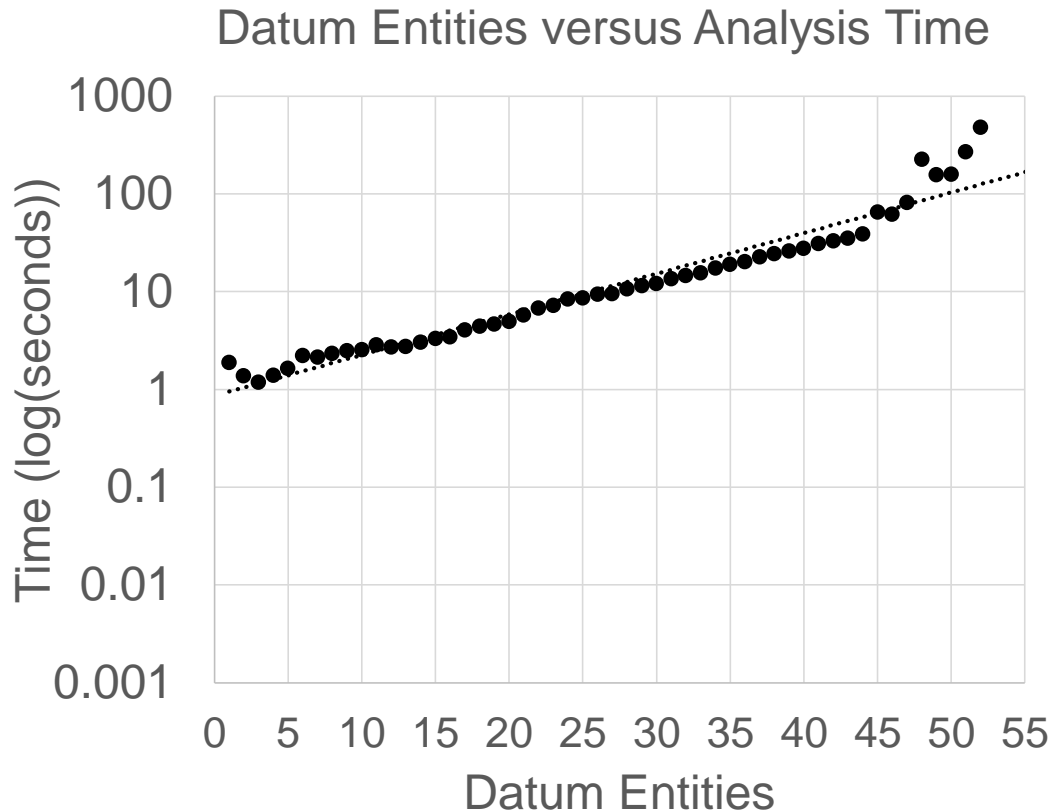
- Most policies have around 50 data requirements.
- Gets bigger when you introduce multiple parties.
- So, how big can we make them before the Eddy toolchain blows up?
  - AKA Does it scale?

## Does it Scale?



- Logarithmic plot.
- How long does it take to do analysis as the number of requirements grows?
- 80: Under 4 minutes.

## Does it Scale?



- New benchmark.
- Logarithmic plot.
- How long does it take to do analysis as the number of data types grow?
- Policy size: fixed, 400.
- 52: Under 8 minutes.

# Conclusions

- Eddy works equally well with multi-party compositions.
- Toolchain scales well to extremely large policies.
- Using two coders and the toolchain, we can analyze a complex compositional system.
- Validate conformance to the 3 privacy principles.
- Two interesting privacy design patterns were found.