FlowCog: Context-aware Semantics Extraction and Analysis of Information Flow Leaks in Android Apps



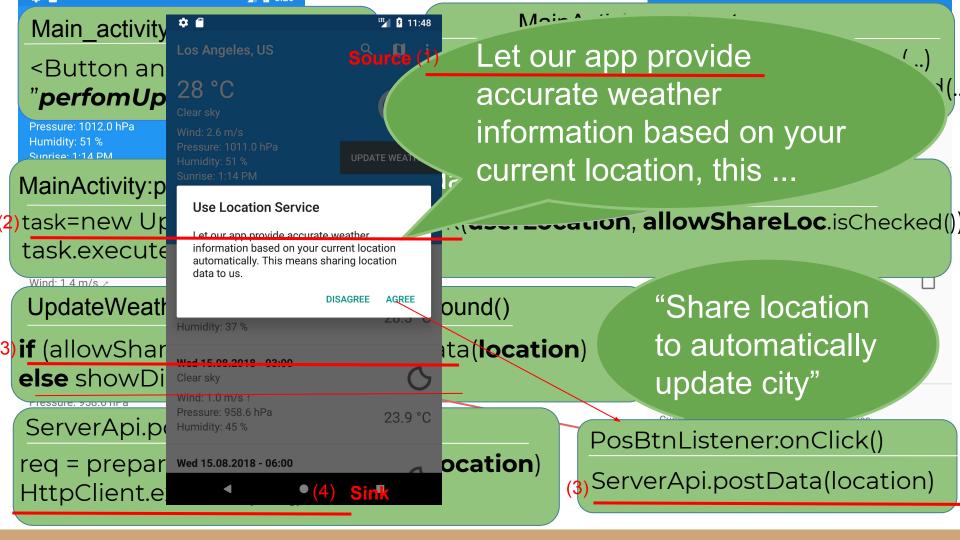
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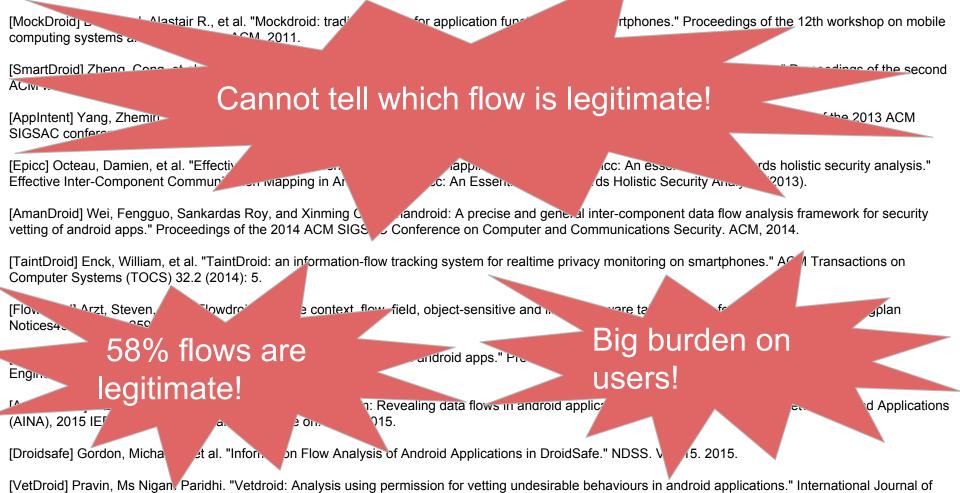


Xiang Pan, Yinzhi Cao, Xuechao Du, Boyuan He, Gan Fang, Yan Chen.

Northwestern University, Johns Hopkins University Zhejiang University, Google

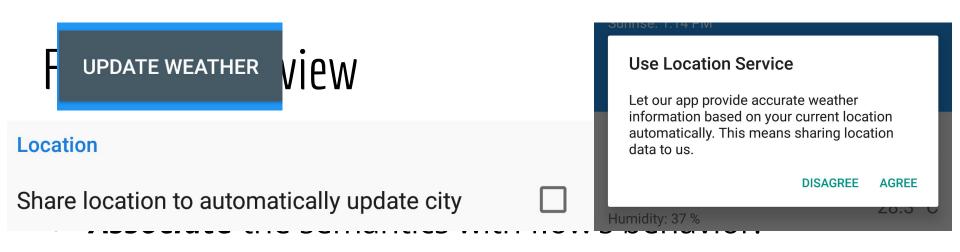
- 1. Motivating Example
- 2. FlowCog Overview
- 3. Design
  - a. View Dependency Explorer
  - b. Flow and Semantics Correlation Inference
- 4. Implementation
- 5. Evaluation & Case Study
- 6. Conclusion



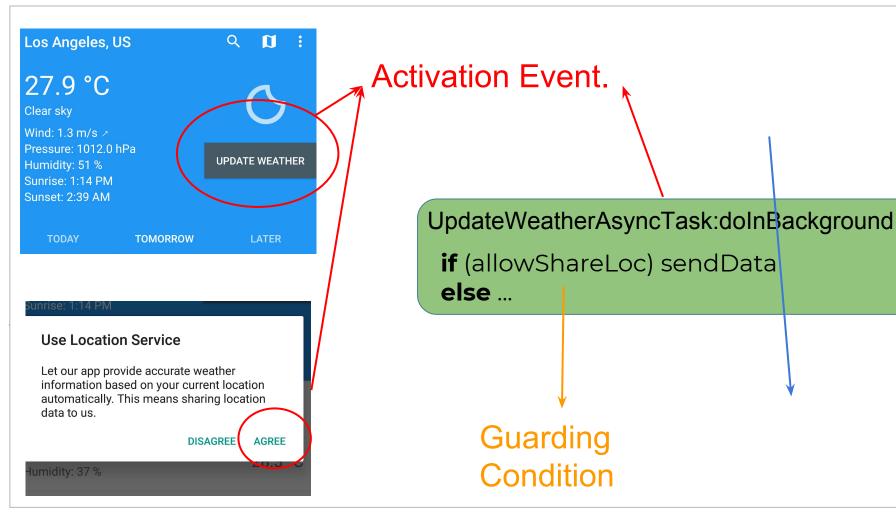


Innovative and Emerging Research in Engineering 2.3 (2015): 131-136.

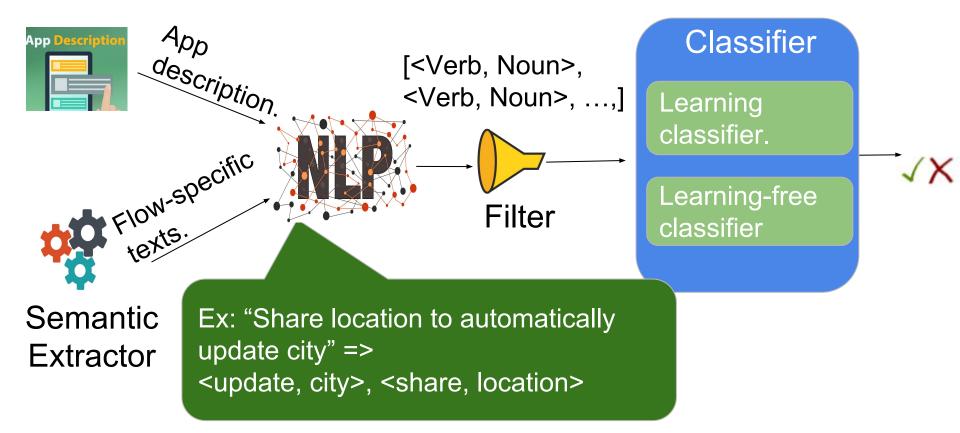
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- High level steps:
  - Associate each flow with its related views via static analysis and an optional dynamic analysis.
  - Extract view semantics. (e.g., "Update Weather")
  - Determine if semantics provides information about flow behavior.



#### FlowCog: Flow and Semantics Correlation Inference (2/2)



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## Design: View Dependency Explorer



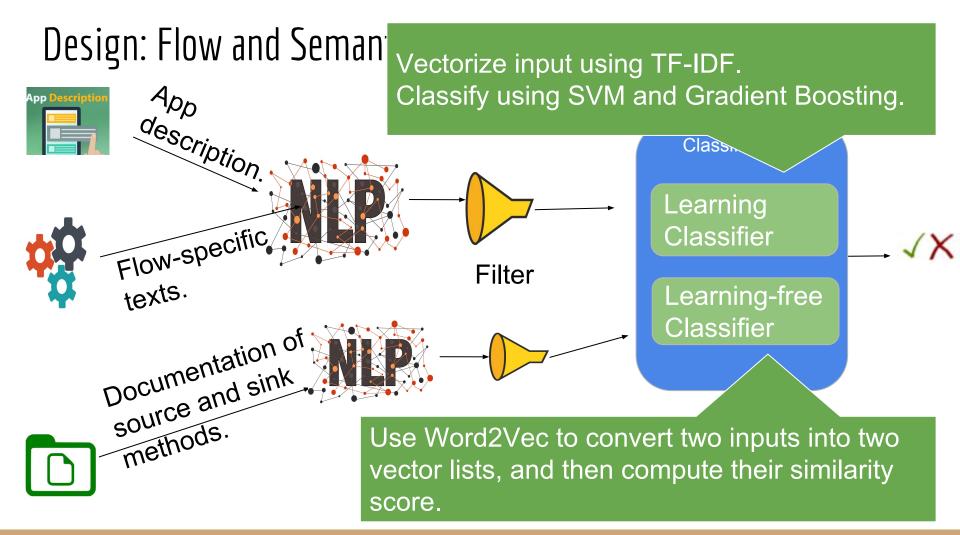
- Sink:
  - Statements in given data flow.
  - Guarding condition statements.

#### All the activation events' registration statements.

Location

Share location to automatically update city

• Use IFDS framework provided by FlowDroid.



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### Implementation

Component	Language	Loc
Flow-related Semantics Extraction	Java	~12,000
Classifier	Python	~3,000
Dynamic Analysis	Python, Java	~1,000
Total	Python, Java	~16,000

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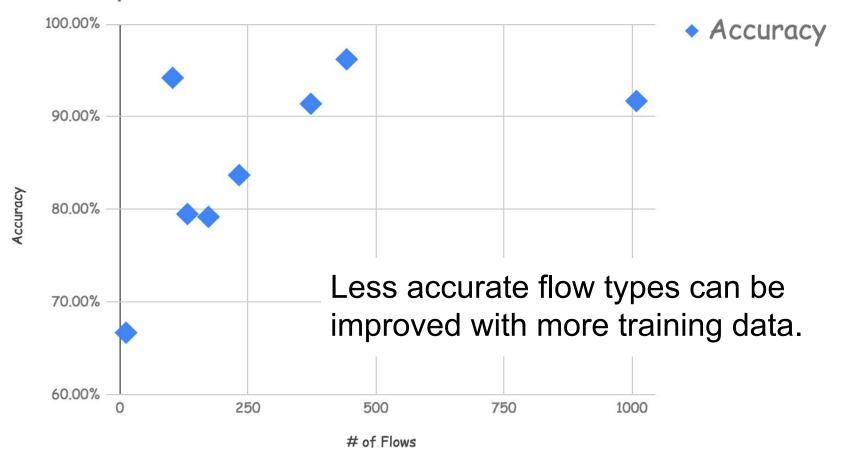
#### Evaluation: Ground Truth

Туре	Apps	Apps with Flows	Legitimate Flows	Maliciou s Flows	Total Flows
Benign	1,299/4,500	361	688	355	1,043
Malicious [Drebin dataset]	586/1,500	255	675	624	1,299
Overall	1,885/6,000	616	1,363	979	2,342

## Evaluation: FlowCog Achieves High Accuracy.

Туре	Flows	Precision	Recall	Accuracy
Benign	1,043	90.3%	95.1%	90.7%
Malicious	1,299	89.9%	91.0%	89.6%
Overall	2,342	90.1%	93.1%	90.2%

#### Accuracy vs. # of Flows



## Case Study: Home of Ocarina

- Leaks out users' geo-location.
- Labeled as legitimate because of extracted semantics.

#### "Map", "The location of home of Ocarina"



## Case Study: SMS Irritate

- Leaks out user-specific information via SMS.
- Labeled as legitimate.

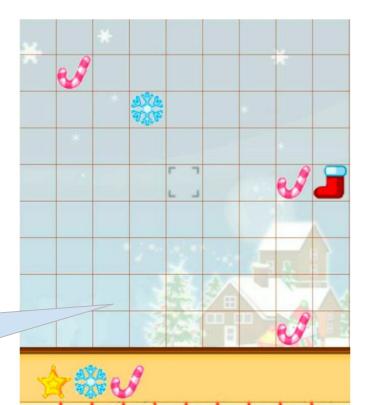
"Send SMS",
"Number of SMS to
flood", "Message"

SMSIrritate	e
Free App: D Send to:	o not buy or sell
Number of S	SMS to flood:
Message:	
	Send SMS
by Sonu	_

## Case Study: Merry Christmas

- Leaks out users' information to Internet.
- Labeled as malicious.

"Move the box to the target empty position..."



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### Conclusion

- FlowCog is **the first system** to extract flow-specific semantics.
- FlowCog adopts NLP techniques to associate flow-specific semantics with flow behaviors.
- Our evaluation results show that FlowCog can achieve a precision of **90.1%** and a recall of **93.1%**.

# Thanks!

#### FlowCog open-source at: https://github.com/SocietyMaster/FlowCog