

# Improving PDG Vector Creation for AnDarwin

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# Background - AnDarwin

AnDarwin project identifies plagiarized Android applications by

- ▶ constructing a *program dependency graph* for each application
- ▶ converting connected components of each PDG into vectors
- ▶ using Locality Sensitive Hashing algorithm to identify clusters of similar vectors

# Background - AnDarwin

## Advantages of this approach

- ▶ avoid solving *maximum common subgraph isomorphism* problem on PDG's, which is known to be NP-hard
- ▶ avoid pairwise comparisons between all  $n$  Android programs in the data set, which would require  $O(n^2)$  comparisons

## Background - PDG

A *program dependency graph*  $G$  is constructed by

- ▶ creating a node for each statement  $s$  in the program
- ▶ for each pair of statements  $s, t$  creating edge  $(s, t)$  if there is a variable in  $t$  whose value depends on statement  $s$

Thus, PDG's are resistant to code reordering, variable renaming and other simple obfuscation techniques.

## Background - PDG Vectors

AnDarwin constructs  $d$ -dimensional PDG vector  $v$  by

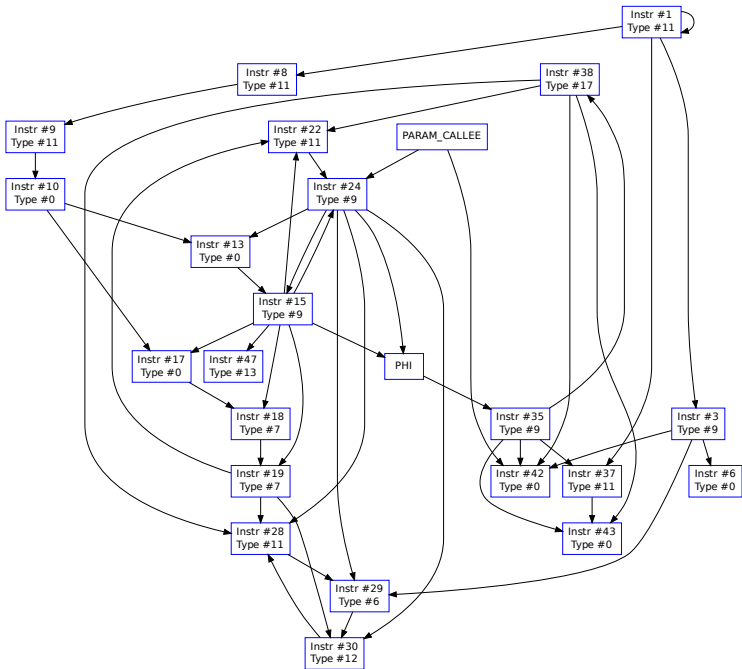
- ▶ classifying program statements into  $d$  types, i.e., conditionals, binary operations, etc.
- ▶ selecting an ordering on the types of statements in the program
- ▶ setting the  $i^{th}$  component of  $v$  to be the number of statements of type  $i$  found in the PDG

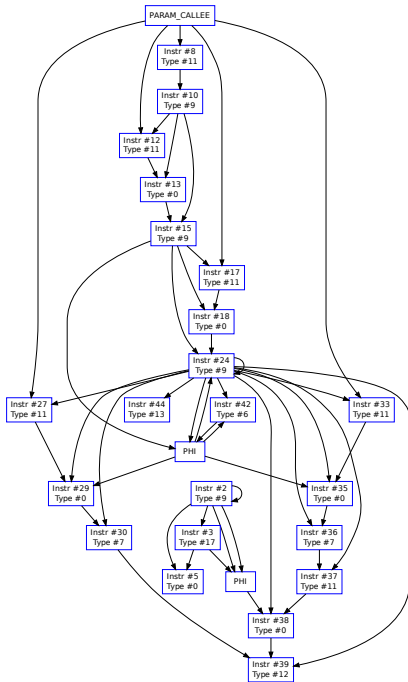
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Unfortunately PDG vectors only encode node count and do not contain any structural information about the graph







# Proposal

Construct a  $2d$ -dimensional PDG vector  $v$  by

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- ▶ setting the  $i^{\text{th}}$  component of  $v$  to be the number of statements of type  $i$  found in the PDG
- ▶ setting the  $(d + i)^{\text{th}}$  component of  $v$  to be the *max out-degree* of the statements of type  $i$

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- ▶ decreases distance between vectors created from programs with different node counts but similar structure



## Advantages - Performance

The LSH algorithm used by AnDarwin has complexity

$$O(d \sum_{g \in G} |g|^p \log |g|)$$

where  $d$  is vector dimension.

Therefore, increasing the vector dimension to  $2d$  only increases the runtime by a constant factor.

# Disadvantages

- ▶ some additional computation time for converting PDG's to vectors
- ▶ since we do not have a characterization for the types of graphs induced by the set of Android applications, this method may potentially create many new false positives

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- ▶ including *average out-degree* in the vector might also be contain useful structural information about the graph
- ▶ implement an automatic method for characterizing false positives

Questions?

