

Deep Dynamic Analysis of Android Applications

By

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THESIS

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To my parents, for instilling in me the determination to follow my dreams.

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to binary fingerprints or system call traces used in other systems. In our tests, we were able to reliably detect three families of malware for which we created signatures with zero false positives.

Secondly, we explore Pyandrazzi’s role in a recent study of advertising fraud on Android, covering over 130,000 Android applications. The system was used to analyze those apps that did not generate ad-related traffic without user interaction. Of the 7,500 apps without such traffic, we found that 12.8% of applications would have generated ad traffic, if they had been properly interacted with via their user interfaces. We then explore augmenting Pyandrazzi to avoid interacting with advertising so that fraudulent behaviors can be better detected. Using a set of rules based on advertising industry standards and common design patterns, we were able to avoid ad-related interactions in 97.6 percent of a test set of 1,000 apps.

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