

A Thesis

entitled

Optimizing Android Memory Management

by Predicting User Behavior

by

Srinivas Muthu

Submitted to the Graduate Faculty as partial fulfillment of the requirements for the  
Masters of Science Degree in Engineering

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December 2015

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An Abstract of  
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With the advent of increase in the amount of RAM (Random Access memory) available in Android smart-phones, there is a case to be made that this additional memory availability can be put to better use. By default, every Android application runs in its own Linux process. When a smart-phone that runs on Android is active, the RAM contains all the active processes and services (processes that run in the background). In addition to these processes and services, the RAM also contains cached background processes. These processes are kept in memory so that in the event the user clicks any of the corresponding applications, they can be loaded onto the screen quickly. If Android needs to reclaim memory for other processes, it eliminates cached background processes through an LRU (Least Recently Used) scheme. We postulate that analyzing user behavior could help us better determine which applications to cache in memory, as cached background processes. We parse the user's Calendar for contextual clues and gather information that could help us predict which application a user is about to use. We measure the cache efficiency (Cache hits and misses, CHR (Cache Hit Ratio)) for the default CRA (Cache Replacement Algorithm) Android employs (LRU), a pure prediction based CRA and finally a hybrid approach that combines the LRU approach with the prediction approach. We also demonstrate why the hybrid CRA is the most efficient one.



To my parents who've always put my needs ahead of theirs.

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I'd like to thank all my friends and family for supporting me throughout my time here at UT but I'd like to especially mention Sandy Stewart for everything she's done for me. This page isn't enough to expound on the details but it suffices to say I wouldn't be here without her and she is like a second mother to me.

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# List of Abbreviations

RAM	.....	Random Access Memory
LRU	.....	Least Recently Used
CHR	.....	Cache Hit Ratio
CRA	.....	Cache Replacement Algorithm
OS	.....	Operating System
AOSP	.....	Android Open Source Project
APK	.....	Android Package
IPC	.....	Inter Process Communication
IC	.....	Integrated Circuits
MB	.....	Mega-Byte
GB	.....	Giga-Byte
APC	.....	Android Process Cache
APMD	.....	Android Powered Mobile Device
CBP	.....	Cached Background Processes
MRU	.....	Most Recently Used
LFU	.....	Least Frequently Used
RR	.....	Random Replacement
OEM	.....	Original Equipment Manufacturer
ADB	.....	Android Debug Bridge
UID	.....	unique Identifier
OOM	.....	Out Of Memory
API	.....	Application Programming Interface
USA	.....	United States of America

# List of Symbols

✓ ..... Represents a check mark indicating that a particular item has been checked or in a different context, whether the checked item is correct

# Preface

This thesis is original, unpublished, independent work by the author, Srinivas Muthu under the tutelage of Dr. Jackson Carvalho.

# Chapter 1

## Introduction

### 1.1 Android Operating System

#### 1.1.1 Overview of the Android OS

Android is a mobile OS (Operating System) based on the Linux kernel and designed primarily for touchscreen devices such as smart-phones and tablets [1]. In addition to touchscreen devices, Android TV, Android Auto and Android Wear are emerging technologies with specialized user interfaces. Globally, it is the most popular mobile OS [2]. Android has an active community of developers and enthusiasts who use the AOSP (Android Open Source Project) source code to develop and distribute their own modified versions of the operating system [4]. Android homescreens are typically made up of app icons and widgets. App icons launch the associated app, whereas widgets display live, auto-updating content such as the weather forecast, the user's email inbox, or a news ticker directly on the homescreen [5].

Internally, Android OS is built on top of a Linux kernel. On top of the Linux kernel, there are the middleware, libraries and APIs written in C and application software running on an application framework. Development of the Linux kernel continues independently of other Android's source code bases.





Figure 1-1: [3] Android System Architecture

## 1.1.2 Applications, Activities and Services

### 1.1.2.1 Applications

Android apps are written in the Java programming language. The Android SDK tools compile the code, along with any data and resource files into an APK (Android Package), which is an archive file with an .apk suffix. One APK file contains all the contents of an Android app and is the file that Android-powered devices use to install the app [6]. The Android operating system is a multi-user Linux system in which each app is a different user. Each process has its own virtual machine (VM), so an app's code runs in isolation from other apps. By default, every app runs in its own Linux process.

### **1.1.2.2 Activities**

An Activity is an application component that provides a screen with which users can interact in order to do something, such as dial the phone, take a photo, send an email, or view a map [7]. An application usually consists of multiple activities that are loosely bound to each other. Typically, one activity in an application is specified as the main activity, which is presented to the user when launching the application for the first time. Each activity can then start another activity in order to perform different actions.

### **1.1.3 Services**

A Service is an application component that can perform long-running operations in the background and does not provide a user interface [8]. Another application component can start a service and it will continue to run in the background even if the user switches to another application. Additionally, a component can bind to a service to interact with it and even perform IPC (Inter Process Communication). For example, a service might handle network transactions, play music, perform file I/O, or interact with a content provider, all from the background.

## **1.2 Growth of available RAM over the years**

### **1.2.1 Advances in Technology**

RAM is a form of computer data storage. A RAM device allows data items to be accessed (read or written) in almost the same amount of time irrespective of the physical location of data inside the memory. The overall goal of using a RAM device is to obtain the highest possible average access performance while minimizing the total cost of the entire memory system. Today, random-access memory takes the

form of IC (Integrated Circuits)(s).

### 1.2.1.1 Moore's Law

Moore's law is the observation that the number of transistors in a dense IC doubles approximately every two years.

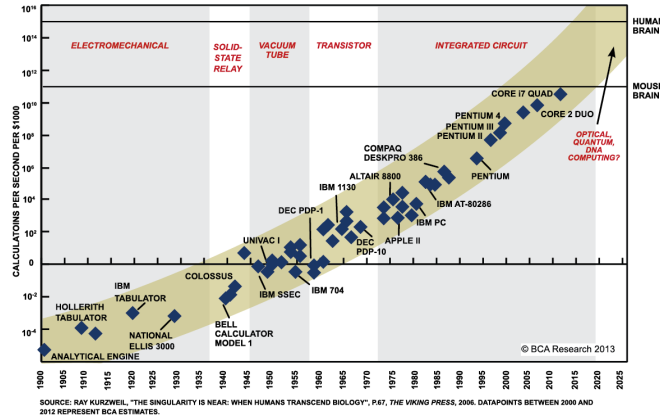


Figure 1-2: [9] Moore's Law

Advancements in digital electronics are strongly linked to Moore's law, especially in the context of memory capacity. T-Mobile G1, the very first Android smart-phone to be released back in 2008 [10] has a 256 MB (Mega-Byte) RAM [11]. In comparison, the Nexus 6P that was released in September 2015 has a 3 GB (Giga-Byte) RAM and some devices like the LG T585 have a 16 GB RAM [12][13], which amounts to 64 times the memory capacity of the T-Mobile G1. The rapid growth in the amount of RAM available to Android devices has in part led to newer possibilities and in our case, a better CRA for determining which processes (in the context of Android applications) remain in memory.

### 1.2.2 Demise of Task Killers

One of the main benefits of the Android OS is the fact that unlike certain other OS(s), it can run apps in the background. This enables us to have multiple applications open at the same time which results in true multitasking. Thus, RAM availability is highly desirable [14]. A popular misconception is that forcibly removing applications from memory in order to 'free up RAM' will result in increased performance. In fact, several task-killer applications promise to do precisely that. With the advancement in the amount of RAM available, the debate should be about how to better use all this extra space, not killing applications to 'free up' more space. In fact, the Android OS can go one step further and pro-actively cache applications that the user might use in the near future. We'll analyze this prospect in more detail in the upcoming chapters.

## 1.3 How Android Manages Processes

### 1.3.1 Android Process Lifecycle

A Android process can be in one of five different states at any given time, from most important to least important [15]:

- Foreground process: A foreground process is one that is required for what the user is currently doing.
- Visible process: A visible process is one holding an Activity that is visible to the user on-screen but not in the foreground.
- Service Process: Service processes are not directly visible to the user, they are generally doing things that the user cares about (such as playing music in the background).

- Background process: A background process is one holding an Activity that is not currently visible to the user. They are kept in an LRU list to ensure the process that was most recently seen by the user is the last to be killed when running low on memory.
- Empty process: An empty process is one that doesn't hold any active application components. The only reason to keep such a process around is as a cache to improve startup time the next time a component of its application needs to run.

Every Android smart-phone user is capable of checking the processes that currently reside in physical memory. The Application Manager which is part of the Settings app shows a list of running processes and cached background processes. Additionally, it breaks down the composition of RAM usage by the System applications and user installed applications.

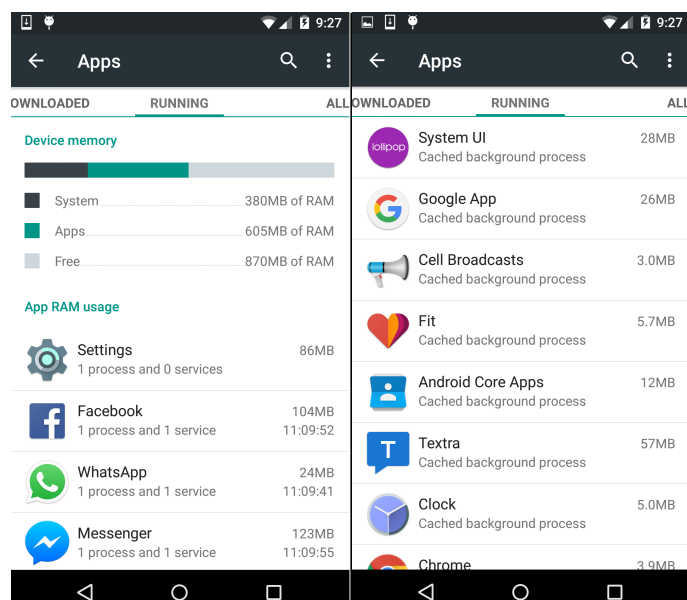


Figure 1-3: Left - List of Running Applications (Active Processes and Services)  
Right - List of Cached Background Processes

### 1.3.2 LRU Cache

LRU is a CRA that discards the least recently used items first. This algorithm requires keeping track of what was used when, which is expensive if one wants to make sure the algorithm always discards the least recently used item [17]. Android keeps processes that are not hosting a foreground app component in a LRU cache. As the system runs low on memory, it may kill processes in the LRU cache beginning with the process least recently used, but also giving some consideration toward which processes are most memory intensive [16].

## 1.4 Goals and Objectives

### 1.4.1 A User-Centric CRA

The main benefit of caching processes (components of applications recently clicked by the user) in memory is to improve startup time (of the application), the next time a component of the application needs to run. This greatly enhances the usability experience and therefore it's in Android's best interests to increase the efficiency of this cache i.e. improve its CHR. In addition to recency of application usage (which is what Android currently relies on in the form of LRU as a CRA), there is scope to potentially infer what application a user may click in the near future. A more user-centric CRA could look into contextual clues and deduce patterns in user behavior to improve the decision making involved in determining which applications to cache in memory, pro-actively or otherwise. We will explore in theoretical detail many such possibilities for user behavior inference and as a proof of concept, read the user's Calendar application and parse for contextual information that could help in improving the CHR of the APC (Android Process Cache).

## 1.5 Organization of Thesis

Several works in the past have focused on context-aware applications and ways to observe user behavior patterns. We'll explore some these works and how this contextual data was put to use in Section 2. In Section 3, we'll take a look at some of the challenges in collecting cache metrics like detecting what's in the RAM, detecting when the user clicks a new application, whether to approach the problem at the user level or the system level, to list a few. Section 4 elaborates on the setup of the experiment and analyzes the design of the applications used to collect these metrics. We dive into the data in Section 5 and summarize its ramifications. We also investigate how certain factors could explain the variation in data. Section 6 talks about scope for future work and explores some of the possibilities of directly adding to this work. We conclude the thesis with Section 7.

# Chapter 2

## Related Work

### 2.1 Early Studies in Context-Aware Computing

Dey, Abowd and Salber, in their 2001 paper titled *A conceptual framework and a toolkit for supporting the rapid prototyping of context-aware applications* laid out the foundation and in many ways a standardized framework for building context-aware applications. When this paper was published, mobile phones were becoming mainstream devices and were no longer confined to the hands of the wealthy. By definition, more people were mobile and the devices they carried with them had the potential to utilize contextual information gathered from the surroundings, in ways that desktops could't do due to their stationary nature. They defined context as any information that characterizes a situation related to the interaction between humans, applications and the surrounding environment. They opined that the state of research in context-awareness was inadequate for three main reasons:

- The notion of context was ill-defined.
- There was a lack of conceptual models and methods to help drive the design of context-aware applications.
- No tools were available to jump-start the development of context-aware appli-



cations.

They focused their efforts on the pieces of context that could be inferred automatically from sensors in a physical environment and produced Context Toolkit [20], a conceptual framework that supported the rapid development of context-aware applications. It is important to keep in mind that at the time this paper was published, the notion of what we refer to as a smart-phone was non-existent. This is evident from the fact that most APMD(s) have built-in sensors that measure motion, orientation, and various environmental conditions. These sensors are capable of providing raw data with high precision and accuracy, and are useful in monitoring three-dimensional device movement or positioning, or changes in the ambient environment near a device [21]. These conditions make APMD(s) very suitable for utilizing contextual information and translating that potential into a better user experience. We utilize this feature of APMD(s) to propose ways to improve user experience by incorporating contextual information in managing the physical memory of the APMD.

## 2.2 Context Patterns in Application Usage

Several works have attempted to collect user behavior data and they've accomplished this in unique ways. Hannu Verkasalo, in his 2007 paper titled *Contextual Patterns in Mobile Service Usage* analyzes how mobile services are used in different contexts [18]. Usage contexts were divided into *home*, *office* and *on the move*. A specialized algorithm tracked user's location patterns over a period of time to determine whether the user was at home or at work. Furthermore, changes in location data revealed whether the user was in transit between home and work. There was a correlation between the mobile services requested by the user and the user's physical location, which enabled the algorithm to understand the user's context quite well (refer Figure 2-1). It was also observed that certain services were requested more

during the weekends as opposed to certain others being requested more during the weekdays i.e. days with office hours.



Figure 2-1: By classifying user context into three distinct entities namely *home*, *work* and *on the move*, the author managed to infer which services were requested more in a given context. For e.g. Voice services i.e. calling was predominantly requested while the user was at home. This type of information can be very useful.

## 2.3 Context Phone

Mika Raento, Antti Oulasvirta, Renaud Petit, and Hannu Toivonen, in their paper titled *ContextPhone: A Prototyping Platform for Context-Aware Mobile Applications* proposed that mobile phones are well suited for context aware computing due to an intimate relationship between the user and the phone [19]. Their primary goal was to provide context as a resource and in order to accomplish that, developed a *ContextPhone* which comprised of four components (Refer Figure 2-2):

- Sensors that acquire contextual data (such as location data from GPS)
- Communication services (such as SMS, MMS to list a few)
- Customizable Applications (that can replace existing applications)
- System Services (such as background services for error logging)



Figure 2-2: The ContextPhone platform; Four interconnected components sensors, system services, communication services, and customizable applications facilitate communication with the outside world.

It is important to observe that this paper came out in 2005, two years before the first version of Android was even launched. Yet, their fundamental point of focus is still relevant in how we can use context as a powerful resource. Their point about the necessity of customizable applications that are needed for context information to be relevant perfectly applies to this thesis. The Android OS which is open sourced as AOSP is as customizable as it can get when it comes to modifying mobile components and by extension the user experience with APMD(Android Powered Mobile Device)(s). To be specific, it facilitates ways to gather cache related information such as the list of processes currently residing in memory, the current foreground application and whether the caching mechanism per se is up for modification to list a few. It would be impossible to collect and/or modify such information on any other mobile OS (such as Apple's iOS).

## 2.4 Using Context Information for Authentication

Dey et al. established a framework for collecting contextual data and Verkasalo and Raento et al. have shown how user behavior can be analyzed through sensors and other modules that gather contextual information. Shi et al.'s paper titled *Implicit Authentication through Learning User Behavior* demonstrates a similar approach in collecting behavioral data of the user but differ in how they put that information to use. They have devised a model to implicitly authenticate smart-phone users based on their behavioral patterns. The user's usage patterns are collected (for e.g. how many calls a user makes a day, how many times user's location changes (Refer Figure 2-3) etc.) and this information is used in building a user model [22].



Figure 2-3: The blue dots represent the users traces in a two-hour epoch over multiple days, and the red ellipses represent the clusters fitted. The major two directional clusters correspond to the users trajectory on a highway

Once the user model is built and the profile completed, the algorithm can determine the likelihood of a random user of the phone not being the owner of the phone (the one whose user model was profiled).

As demonstrated by Shi et al., there are numerous ways contextual information can come in handy. In our case, we propose a user-centric CRA for the APC, demonstrate

its superior efficiency and user experience, by looking into the user's calendar for contextual information.

# Chapter 3

## Challenges in Collecting the Desired Metrics

### 3.1 Introduction

Now that we've laid the foundation for using contextual information to influence which processes Android caches in memory, we need to establish the metrics that need to be collected in order to measure the efficiency of the existing CRA and the proposed CRA. The following section addresses the metrics we need.

### 3.2 What are the desired metrics?

#### 3.2.1 Hits, Misses, Hit Ratio & Latency

Alan Jay Smith defines Cache as a high speed buffer that holds items in current use [23]. In our context, Android holds application components (such as processes) as CBP (Cached Background Process)(s) in memory. As previously discussed, CBP(s) do not take up processor time, they are merely cached for quick application startup should the user request that particular application. A cache hit is a state in which data requested for processing by a component or application is found in the cache

memory. It is a faster means of delivering data to the processor, as the cache already contains the requested data [24]. When applied to our scenario, a cache hit would represent a situation wherein the user requests (clicks) an application that currently resides in memory, either as an active process (including services) or as a CBP (Refer Figure 3-1).

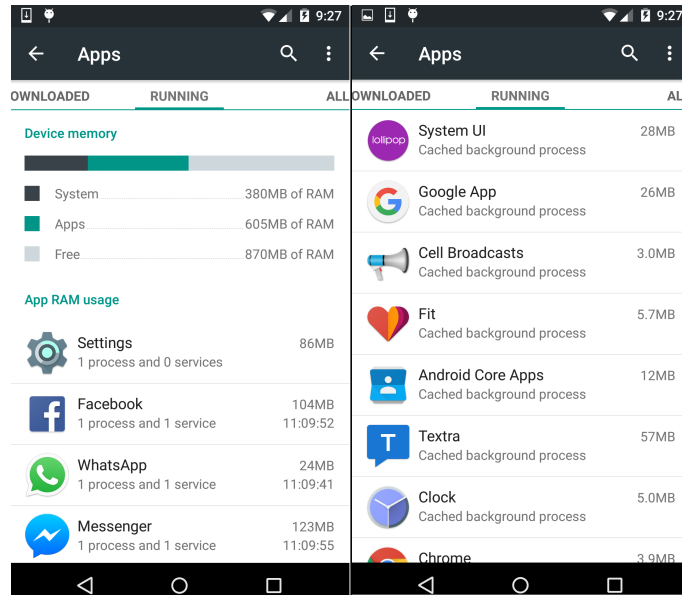


Figure 3-1: Given this snapshot of the RAM, if the user clicks on the Messenger application (active process) or Clock application (CBP), it would result in a Cache Hit

A cache miss is a state where the data requested for processing, by a component or application is not found in the cache memory [25]. When applied to our scenario, a cache miss would represent a situation wherein the user requests (clicks) an application that currently does not reside in memory, either as an active process (including services) or as a CBP (Refer Figure 3-2).

CHR is defined as the percentage of cache hits compared to the overall number of requests (total of cache hits and cache misses).



Figure 3-2: Given this snapshot of the RAM, if the user clicks on say, the Google Drive application (process not in memory), it would result in a Cache Miss

$$OverallNumberOfRequests = CacheHits + CacheMisses \quad (3.1)$$

$$CHR = \frac{CacheHits}{OverallNumberOfRequests} * 100 \quad (3.2)$$

There are two primary figures of merit of a cache [23]:

- Latency
- CHR

The latency of a cache describes how long after requesting a desired item, the cache can return that item (in the case of a cache hit). In our scenario, processes are cached in RAM. Therefore, even if the number of processes cached in memory increases, there will hardly be any difference in the time it takes to fetch a requested item (application component) as they all reside in RAM which by definition supports



random access. Latency is more of a factor in caches that are located further away from physical memory (such as L1 and L2 caches).

The CHR of a cache describes how often a searched-for item is actually found in the cache. The higher the CHR, the better the cache is. When applied to our context, a high CHR results in more application components being fetched from memory, rather than disk which in turn results in quicker application startup times. This improvement in startup speed results in an enhanced user experience.

### **3.2.2 CRA(s) Under Consideration**

Now that we've determined the fundamental metrics we need, let's analyze the various CRA(s) including the one Android currently uses, namely LRU.

#### **3.2.2.1 Optimal CRA**

The most efficient CRA would be to always discard the information that will not be needed for the longest time in the future. In our case, the ideal CRA would be one that ensures that every application the user requests, resides in memory. This optimal result is referred to as Belady's optimal algorithm [26] or the clairvoyant algorithm. Since it is generally impossible to predict how far in the future information will be needed, this is not implementable in practice. The practical minimum can be calculated only after experimentation, and one can compare the effectiveness of the actually chosen CRA against Belady's as a benchmark.

#### **3.2.2.2 LRU**

This is the default CRA Android currently uses to discard processes from memory. LRU is a type of CRA that discards the least recently used items first. This algorithm keeps track of what was used when, and it's expensive to ensure that the least recently used item is always discarded first. When applied to our scenario, Android caches

application components in memory as CBP(s) and associates each with a timestamp. If the memory gets too full, it starts removing the least frequently used CBP(s) first and depending on how direly low the memory is, it may not stop killing processes until even the active foreground application (the one user is interacting with) is removed, [15] although this situation is extremely rare in practice and only results in the case of malfunctioning applications that leak memory. The Settings application can give the user clues about which CBP(s) were recently used. Clicking on an application that is currently not in memory results in that application being cached as a CBP near the top of the list. This suggests that the CBP(s) displayed by the Settings application are roughly sorted by their timestamp from MRU to LRU (Refer Figure 3-3).

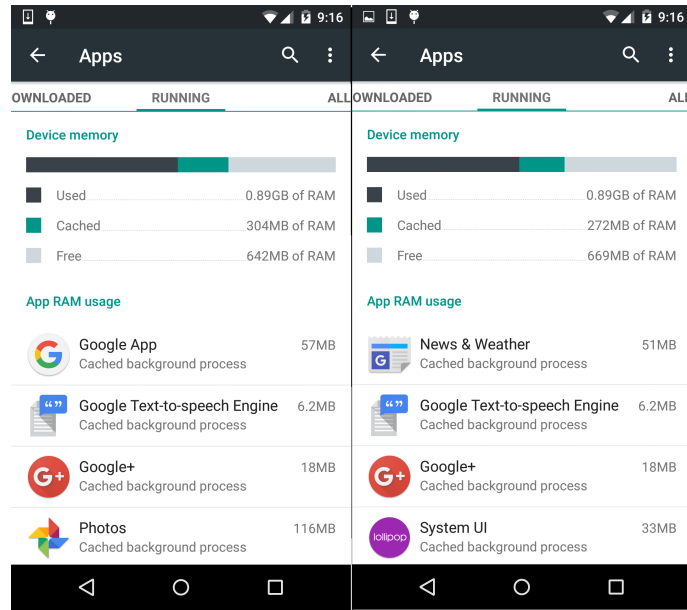


Figure 3-3: The image to the left describes the snapshot of the CBP(s) as displayed to the end-user by the Settings app. The image to the right displays the snapshot of the CBP(s) after the end-user requests for the News and Weather application. Note that the News and Weather application will not be cached as a CBP until the user is done interacting with it as up until that point, it will reside in memory as an active process.

It is worth noting that a perfect implementation of LRU requires a timestamp on

each reference, and the OS needs to keep a list of items ordered by the timestamp. This implementation may be deemed too expensive in certain cases especially if the frequency of memory references is high. The common practice is to approximate the LRU behavior and this can be achieved in other ways [27].

### **3.2.2.3 Most Recently Used (MRU)**

MRU discards, in contrast to LRU, the most recently used items first. In findings presented at the 11th VLDB conference, Chou and DeWitt noted that when a file is being repeatedly scanned in a looping sequential reference pattern, MRU was the best replacement algorithm [28]. Subsequently other researchers presenting at the 22nd VLDB conference noted that for random access patterns and repeated scans over large datasets (sometimes known as cyclic access patterns) MRU cache algorithms had more hits than LRU due to their tendency to retain older data [29]. In our scenario, CBP(s) neither comprise a 'large dataset' nor are they repeatedly scanned in a looping sequential reference pattern (as they are only invoked during application requests which are on average, not that often), therefore it's unlikely that MRU would result in a high CHR but it's worth mentioning nonetheless. MRU algorithms are most useful in situations where the older an item is, the more likely it is to be accessed, which again is not a feature of CBP(s).

### **3.2.2.4 Least Frequently Used (LFU)**

LFU is a type of CRA usually associated with memory management within a computer. The standard characteristics of this method involve the system keeping track of the number of times an item is referenced in memory. When the cache is full and requires more room the system will purge the item with the lowest reference frequency. This could be a really good fit with our scenario given the tendency of most smart-phone users to stick with the same set of popular applications and

use them in rotation [31]. Under these circumstances, the frequency of a select few applications would be really high and would always be cached in memory as they're requested often. LFU is sometimes combined with a Least Recently Used algorithm called LRFU [30]. We'll explore some of the possibilities with LFU in the future scope section and in our proposed hybrid approach where we incorporate global frequency of application usage in obtaining contextual information.

#### **3.2.2.5 Random Replacement (RR)**

Much like the name suggests RR is a CRA that randomly selects a candidate item and discards it to make space when necessary. This algorithm does not require keeping any information about the access history. For its simplicity, it has been used in ARM processors [32]. When applied to our scenario, it doesn't seem like a viable option, mainly because there isn't a benefit in randomly caching applications in memory.

#### **3.2.2.6 LRU - Context Hybrid**

So far we've analyzed the standard CRA(s) that focus on discarding items that were either accessed least recently, least frequently or in other standard ways. Our approach combines the applications that Android caches through an LRU scheme with a separate context-analyzer module that reads the user's calendar and predicts which applications the user is likely to use. When the user requests an application, the combined set of applications (default applications in LRU and the list of suggested applications) is used as the base to predict whether a cache hit or cache miss occurred. We'll analyze in detail how each module functions but before that, let's take a look at whether Android caches CBP(s) pro-actively and how applications will be removed from memory in the cases where it runs too low.

**Pro-actively Adding CBP(s)** In order for the proposed LRU-Context Hybrid (henceforth referred to as the Hybrid approach) CRA to work, there must be provision for the Android OS to pro-actively cache application components in memory as CBP(s). Presuming that the context-analyzer module does its job and suggests a list of applications the user might use in the near future (based on user's Calendar information), there must a mechanism for Android to utilize that information and pro-actively cache these applications in RAM as CBP(s). This mechanism can be verified at the user level through a small experiment. As previously mentioned, the Settings application provides a visual interface for the list of active processes and services running in memory and the CBP(s) that reside in memory at any moment in time. As part of the experiment, each and every CBP was forcefully removed from the cache (there is provision to do this at the user level (Refer Figure 3-4)).

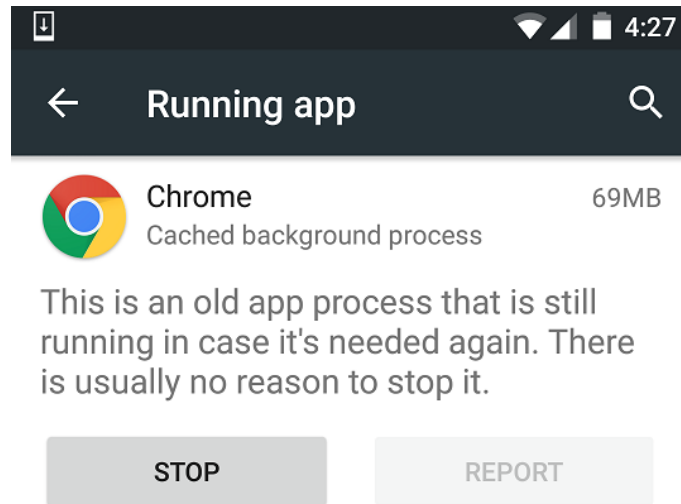


Figure 3-4: Pressing the *STOP* button removes the CBP from memory

Once each CBP was removed, the phone was untouched for three minutes. After three minutes, when the Settings application was launched to look at the list of CBP(s), it wasn't empty. Instead, there were several CBP(s), some of which were previously removed and some of which were't in the list to begin with (Refer Figure

3-5).

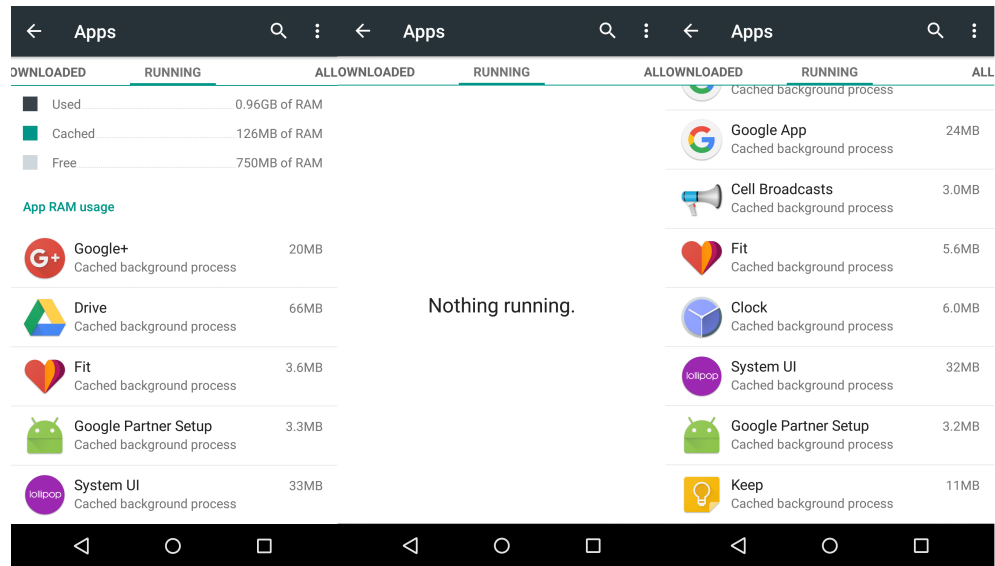


Figure 3-5: The leftmost image shows the snapshot of the CBP(s) before forceful deletion. After each and every CBP was removed manually, the screen displayed is shown in the middle image. The image farthest to the right shows the repopulated list of CBP(s), 3 minutes after the middle image was taken. This demonstrates Android's ability to utilize free RAM space and pro-actively cache CBP(s)

This suggests that Android recognizes that an under-utilized RAM is a waste of resource and pro-actively caches applications that were'nt in memory to begin with, as CBP(s). This is also the reason why task killer applications fell out of favor [33]. This demonstrates that not only does Android provide a mechanism to pro-actively cache applications in memory, it already does so. The applications suggested by the Hybrid CRA approach can be introduced into the memory as CBP(s) in the same way. This is discussed in future scope section.

**Prioritizing Overall Cache** The Hybrid CRA would involve storing as CBP(s), both default application components cached by Android's LRU scheme and the application components of the list of applications obtained from analyzing the user's

Calendar. So the question now becomes, who gets kicked out when memory is running low? Since the applications inferred from user’s contextual information do not have a recency associated with them, the LRU CRA cannot be applied over the overall list of CBP(s). A simple solution is to prioritize the CBP(s) obtained from the list of suggested applications over the default LRU based CBP(s) i.e. all the default CBP(s) are removed from memory before the CBP(s) cached as a result of context inference. There are two advantages to this approach:

- Only few applications are suggested by the context module.
  - For instance, say the user has a Calendar entry that reads *Call Mom*. There will be two or at most three applications suggested by the context-module that would be potentially useful to the smart-phone user in the context of calling his mother. This makes it more convenient to target CBP(s) cached by the LRU scheme in case of low memory.
- The probability of user requesting an application related to the Calendar event is quite high.
  - This is due to the fact that a recorded event in a user’s Calendar is highly likely to occur and the applications needed in this context have a higher chance of being utilized as opposed to CBP(s) cached by the LRU scheme.
  - This is explored further in the future scope section where we discuss the relative probabilities of applications being used in conjunction with a Calendar event.

### 3.2.3 Supplementary Data

We’ve talked about the fundamental metrics involved in measuring the efficiency of the APC, namely cache hits, cache misses and CHR. We explored potential candidates

for serving as the CRA in managing the APC. There are other subtle measurements that aren't as significant as the fundamental metrics but are important nonetheless as they shed light on certain attributes of the experiment. They are listed below:

- Number of Installed Applications
- Number of Unique Applications Requested (Clicked)
- Phone Model
- Android OS Version

Depending on the value of the overall number of installed applications and the overall number of unique applications requested by the end-user, certain CRA(s) have an edge over the others. Phone model information is collected to see if there is any correlation between models and cache efficiencies. Some of the techniques used to gather context information are not possible in certain Android OS versions. We'll analyze the correlation between these fields and the efficiency of CRA(s) in the Section 5.

### **3.3 Challenges**

Now that we've established what the desired metrics are, let's examine some of the challenges in collecting them.

#### **3.3.1 Deciding Between System vs User Level Approach**

Firstly, we need to determine whether to approach the problem from a user level perspective or not. Attempting to solve the problem i.e. collecting the aforementioned metrics at the user level would involve building Android applications to gather relevant data. On the other hand, approaching the problem from a system level



would involve altering the source code of the Android OS (facilitated by AOSP) and recompiling it to build our own custom distribution, much like the CyanogenMod community [34]. It is easier to solve the problem (if possible) at the user level for two reasons:

- The complexity involved in altering OS source code is significantly higher compared to writing Android applications.
- It would be significantly harder to convince potential research volunteers to reboot their smart-phones with a custom distribution of the Android OS as opposed to asking them to install an Android application or two.

Given that it's beneficial to gather relevant data at the user level, we need to identify the complexities involved and determine whether they are solvable at the user level. In order to effectively determine which CRA is ideal for the APC, we need to identify whether the following problems can be approached from a user level perspective:

- Getting the list of processes currently in memory ✓
- Knowing when a new process is launched by the user ✓
- Reading the User's Calendar ✓

We'll inspect in detail, why and how we need to solve these problems (and others) in the next section but the '✓' indicates that they are indeed solvable at the user level and do not require system level changes.

### **3.3.2 Getting Processes in Memory**

Let's first address why it's necessary to get the list of processes that reside in memory. In our scenario, we are trying to determine the most efficient caching scheme amongst the following:

- LRU (Default)
- Pure Prediction (Applications will be cached solely based on context inference)
- LRU-Context Hybrid (or simply Hybrid)

In order to determine which scheme is most efficient, we need the ability to measure CHR data and by extension if and when a cache hit or miss occurs. We can only determine if a cache hit occurred if we had access to the list of processes in the cache i.e. the active processes and the CBP(s) in memory. The *ActivityManager* class in Android provides two relevant helper methods [35]:

- *getRunningTasks()*
- *getRunningAppProcesses()*

Unfortunately, as of Android 5.0, it has become increasingly difficult to get a list of applications running in memory. *getRunningTasks()* has been deprecated and *getRunningAppProcesses()* has been killed (It only returns the requesting application rather than all applications running in memory). The documentation has't changed and Google is ignoring requests to either update the documentation or restore the original implementation. An alternate solution is to use the *UsageStatsManager* class [36]. However, it requires special permission from the user in the Settings application and some OEM (Original Equipment Manufacturers)(s) have removed this setting explicitly rendering this solution unreliable. The best solution involves using the ADB (Android Debug Bridge) shell to retrieve the desired information which can then be parsed into a suitable list of processes residing in memory. ADB is a versatile command line tool that enables communication with an emulator instance or a connected APMD. It is a client-server program that includes three components [37]:

- A client, which runs on the APMD. The client can be invoked from a shell by issuing an ADB command. Other Android tools can create ADB clients as well.

- A server, which runs as a background process on the APMD. The server manages communication between the client and the ADB daemon running on the APMD.
- A daemon, which runs as a background process on the APMD.

The ADB provides a Unix shell that can be used to run a variety of commands on the APMD. The command binaries are stored in the APMD's file system at */system/bin/* [38]. We eventually obtain the list of processes residing in memory by running the *toolbox* command [42] in the shell. The Android *toolbox* command encapsulates the functionality of many common Linux commands (and some special Android commands) into a single binary [39]. Some of the commands provided by Toolbox are listed below (not an exhaustive list):

- cat, chmod, chown, cmp, cp
- ctrlaltdel, date, dd, df, dmesg
- toolbox, top, touch, umount
- ps, reboot, renice, rm, rmdir

Of these, we are most concerned with the *toolbox* and *ps* commands as they provide a direct window into the processes residing in memory. We use the following parameters with the *ps* command to get the desired process data:

- '-P': It shows the scheduling policy.
- '-p': It displays the priorities and niceness level.
- '-x': It shows system time in seconds
- '-c': It displays the CPU

With the help of *libsuperuser* library which provided a convenient framework to run shell commands [40] and Jared Rummler’s *AndroidProcess* library that provided a framework to parse the process-related shell outputs [41], the following algorithm (Refer Algorithm 1) was written to obtain a list of processes residing in memory (active processes and CBP(s)). It is worth noting that this list corresponds to the list of processes displayed by the Settings application (Refer Figure 3-1) as part of the running applications and CBP screens, and that the Settings application has system privileges that third party Android applications don’t.

**Result:** LIST Process

INITIALIZE LIST Process - processes

LIST String - stdout — Shell.SH.run("toolbox ps -p -P -x -c")

Integer myPid — android.os.Process.myPid()

```
for String line : stdout do
|   INITIALIZE Process - process
|   if process matches APP-ID-PATTERN then
|   |   if (process.ppid EQUAL TO myPid)
|   |   |   OR (process.name EQUAL TO "toolbox") then
|   |   |   |   CONTINUE
|   |   else
|   |   |   ADD TO processes - process
|   |   end
|   else
|   |   CONTINUE
|   end
end
```

RETURN processes;

**Algorithm 1:** In the above algorithm, *Process* is a custom data structure that parses the output format of the *toolbox ps* command. Its class structure is discussed in detail in Section 4.4. The *APP-ID-PATTERN* is a custom regular expression that maps to application ID patterns for the Android OS in versions 4.0 and above. Each process is added to a result list that is returned by the algorithm. Note that the requesting application represented by *myPid* (discussed in 4.4) and *toolbox* related processes represented by process names that match *toolbox*, aren't added to the result.

### 3.3.3 Obtaining the Foreground Application

In order to update the CHR metric, we need the ability to detect new cache hits and cache misses. A new cache hit/miss occurs when there are new application requests i.e. changes in the foreground application. Before we analyze how to detect change in the foreground application, we first need a way to obtain the current foreground application. Had we approached this problem at the system level, we could've made use of system level privileges such as listening to a new screen-touch by the user and detecting which application was launched but this is not possible from a third party user-level application. Fortunately, there is a workaround to detecting the foreground application at the user level. We obtain every process currently running in memory and through a process of elimination, obtain the foreground application. To elaborate, the foreground application has certain unique properties in its process name, UID (Unique Identifier) and other fields, that can be exploited to eliminate processes that are not associated with the current foreground application. The following algorithm (Refer Algorithm 2) illustrates this process.

**Result:** Foreground Application

INITIALIZE LIST *files* — LIST */proc* File

**for** *File file : files* **do**

**if** *file IS A DIRECTORY* **then**

        | CONTINUE

**end**

*pid* — *file.getName()*

*cgroup* — FILE READ (*/proc/%d/cgroup*)

*cpuSubSystem*, *cpuAccountSubSystem* — *cgroup*

**if** *cpuAccountSubSystem ENDS WITH pid OR cpuSubSystem ENDS*

*WITH bg\_non\_interactive* **then**

            | CONTINUE

**end**

*cmdline* — FILE READ (*/proc/%d/cmdline*)

*uid* — PARSE FROM *cpuAccountSubSystem*

**if** *cmdline CONTAINS com.android.systemui OR uid BETWEEN 1000*

*AND 1038* **then**

            | CONTINUE

**end**

*oomAdj* — FILE READ (*/proc/%d/oom\_score\_adj*)

*oomScore* — FILE READ (*/proc/%d/oom\_score*)

**if** *LOWEST oomScore AND oomAdj EQUAL TO 0* **then**

        | RETURN *cmdline*

**end**

**end**

**Algorithm 2:** This algorithm obtains the current foreground process i.e.

the process associated with the application that the user is  
interacting with.

Firstly, all files with path */proc* are obtained and each iterated over, to find the foreground application. We eliminate file directories as these are of no concern to us. If the CPU account sub-system ends with *pid*, it's not an application process and is thus eliminated. Similarly, if the CPU sub-system ends with *bg\_non\_interactive*, it's a background process and is eliminated as well. The application stored in *cmdline* is a potential candidate for the foreground application and is the right choice as long as it clears four more hurdles:

- *cmdline* does not contain *com.android.systemui*
- *uid* does not lie between 1000 and 1038
- *oomAdj* i.e. the OOM (Out Of Memory) adjustment level [43] equals 0
- *oomScore* is the lowest among all iterated values

If *cmdline* contains *com.android.systemui* or if *uid* lies between 1000 and 1038, it's a system application and is thus eliminated from consideration. The crucial attribute of the foreground application is that it has the lowest OOM score of all applications. When the Android OS is running too low on memory, it is the job of the Linux OOM killer [44] to sacrifice one or more processes in order to free up memory for the system when all else fails. A low OOM score indicates that the process is less likely to be killed and that it is lower on the OOM killer's radar. It quite intuitive that the process with the lowest OOM score would be the foreground process as it is of paramount importance not to remove the process, the end-user is interacting with, unless it's the very last resort. This unique property of the foreground process enables us to distinguish it from the rest of the pool.



### 3.3.4 Updating the Statistics

So far, we’ve established ways to collect the list of processes in memory and the foreground application. The next step is to figure out a way to update the metrics as and when the user requests new applications. The rate at which the end-user toggles between applications varies from person to person. It also depends on the level of usage. For instance, there are times when the end-user is actively using their APMD and others when the APMD is idle or is switched off. Therefore it’s hard to predict when there could be a potential change in the foreground application. There is no inherent mechanism for user-level applications to detect third party application launches (That privilege is reserved for system level processes). Therefore, our solution comprises of starting a service [8] that retrieves the current foreground process (Refer Algorithm 2) and the current list of processes residing in memory (Refer Algorithm 1) every second. If there is a change detected in the foreground process (compared to the last second), then the cache metrics are updated. The time interval of one second is chosen as it deemed to be reasonably accommodative of the average time it takes an end-user to switch from one application to another. This is further discussed in the Points of Weakness section. The design of the service and the corresponding application is discussed in sections 4.4 and 4.5. The following algorithm (Refer Algorithm 3) illustrates the process of updating the cache metrics.

**Result:** Update Cache Metrics

```
for EACH SECOND do
    newForegroundApp — getNewForegroundApp()
    currentListOfProc — getListOfProcesses()
    if newForegroundApp EQUALS oldForegroundApp then
        | CONTINUE
    end
    if oldListOfProc CONTAINS newForegroundApp then
        | cacheHit++
    else
        | cacheMiss++
    end
    updateMetrics()
    oldListOfProc — newListOfProc
    oldForegroundApp — newForegroundApp
end
```

**Algorithm 3:** This algorithm retrieves the foreground application and checks if it has changed since the last second. If it has, it verifies whether it was a cache hit or a cache miss and updates the cache metrics. Finally, it updates the old foreground application and the previous list of processes to the current one so that the next iteration uses these values as the base.

### 3.3.5 Backing Up the Data

Previously, we discussed the mechanism by which the background service updates the cache metrics. The other aspect of this issue is storing this data. Before we inspect how to store the data, let's discuss the necessity for such a provision. Our background service is designed to run throughout the duration of the experiment and retrieve the foreground application and the list of processes in memory every second.

This results in it's OOM score increasing i.e. it's likelihood of getting killed by the Linux OOM killer [44] which would result in a loss of data. It is worth noting that the OOM killer selects the *best* process to kill which is decided by the combination of how long a process has been running and how much memory would be released if the process is killed. Even though our process scores poorly on the time duration aspect, it takes up relatively low memory compared to some of the other common background processes (Refer Figure 3-6).

Android provides several options for saving persistent application data. There are five ways to accomplish this task [45]:

- Shared Preferences - It is used in cases where primitive data needs to be stored in Key Value pairs.
- Internal Storage - It is used to store private data on device memory.
- External Storage - It is used to store public data on shared external storage
- SQLite Databases - They are used to store structured data in a relational database.
- Network Connection - It is used to store data over the Internet.

For our scenario, Internal Storage is the best fit as it keeps the data private to the application and since the data is quite primitive (cache hits and misses), we do not need SQLite databases or network connections. The data is written to a file every minute. That way, we don't have to constantly perform I/O operations from the service and in the worst case scenario, fifty nine seconds of Cache data would be lost which is relatively a small amount of time. Algorithm 3 can be tweaked to incorporate the file write operation (Refer Algorithm 4).

**Result:** Update Cache Metrics AND Write to File

```
for EACH SECOND do  
  ...  
  ...  
  updateMetrics()  
  if END OF MINUTE then  
    | writeMetricsToFile()  
  end  
  ...  
  ...  
end
```

**Algorithm 4:** In addition to updating the metrics every second, it also writes the data to internal storage every minute. Any time the application is closed and launched again the metrics are initialized with the values in file.

### 3.3.6 Reading the User's Calendar

To compute the CHR of the Hybrid CRA approach, we need the ability to obtain the list of events from the user's calendar and the ability to parse that information to come up with a list of suggested applications. We'll discuss the parsing aspect in the next segment and focus on finding a way to read the user's calendar. The solution to this problem is quite simple, Google provides a public API (Application Programming Interface) [46] to obtain access to the user's calendar information, including the list of events. This operation doesn't have to be carried out with the same frequency as reading the list of processes in memory or obtaining the foreground application. In our solution, we query the calendar every four hours (for events occurring in that four hour period) and feed the data (events list) into a module that parses the information and produces a suggested list of applications (Refer Algorithm 5).

**Result:** Update Cache Metrics, Write to File AND Update List of Suggested Applications

```
for EACH SECOND do
  ...
  ...
  updateMetrics()
  if END OF MINUTE then
    | writeMetricsToFile()
  end
  if END OF 4 HOURS then
    | updateListOfSuggestedApplications()
  end
  ...
  ...
end
```

**Algorithm 5:** In addition to updating the metrics every second and writing them to file every minute, it also updates the list of applications suggested by the Calendar Parser (discussed in next segment)

The following algorithm (Algorithm 6) illustrates our implementation of the module that queries the user's calendar.

**Result:** Get List of Events from Calendar

INITIALIZE List *result*

List *allEvents* — QUERY *content://com.android.calendar/events*

**for** *EACH event IN allEvents* **do**

**if** *event.startTime()* IS WITHIN 4 hours FROM NOW **then**

        | *result.add(event)*

**end**

**end**

RETURN *result*

**Algorithm 6:** This algorithm obtains all events in the user’s calendar that start within 4 hours from the time of the algorithm invocation. Since this algorithm is invoked every four hours, every event in the user’s calendar is parsed although it’s worth noting that once read, a change in the event details is not translated to the application. This is discussed further in the Points of Weaknesses section.

### 3.3.7 Parsing the Calendar Information

We’ve examined the mechanism behind reading the user’s calendar, thereby solving one half of the puzzle. The other half is utilizing this information to compute the Hybrid CRA’s metrics. In order to achieve that, we need to build a module that parses the list of events obtained from reading the calendar and produces a suggested list of applications. Every event in the user’s calendar is split into individual keywords and a predetermined mapping between keywords and applications builds the list of suggested applications. For instance the keyword *mail* is associated with the *GMail* application, among others. This mapping is determined by a combination of three things:

- The suggestions from the Google Play Store [48] when searched with the key-

word.

- The number of downloads among competing applications for the same keyword.
- Popularity of the applications in United States of America (USA) [47].

To elaborate, when the Play Store was searched with the keyword *skype*, Skype application was the top result. Once it is verified that Skype is part of the installed applications [49], it is added to the list of suggested applications. It is worth noting that most of the suggested applications are present by default on Android (such as GMail, Hangouts, Docs to list a few) and the only applications that are checked against the list of installed applications are highly popular applications [47]. Additionally, if there are no events in the user's calendar for the four hour period, then the list of suggested applications is populated with the ten most popular Android applications in USA [47]. Theoretically it seems likely that the user might use the applications suggested by Google's application search engine, for a particular keyword. However, it remains to be seen whether the suggested applications result in cache hits and by extension a better caching scheme. We'll analyze this data in Chapter 5.

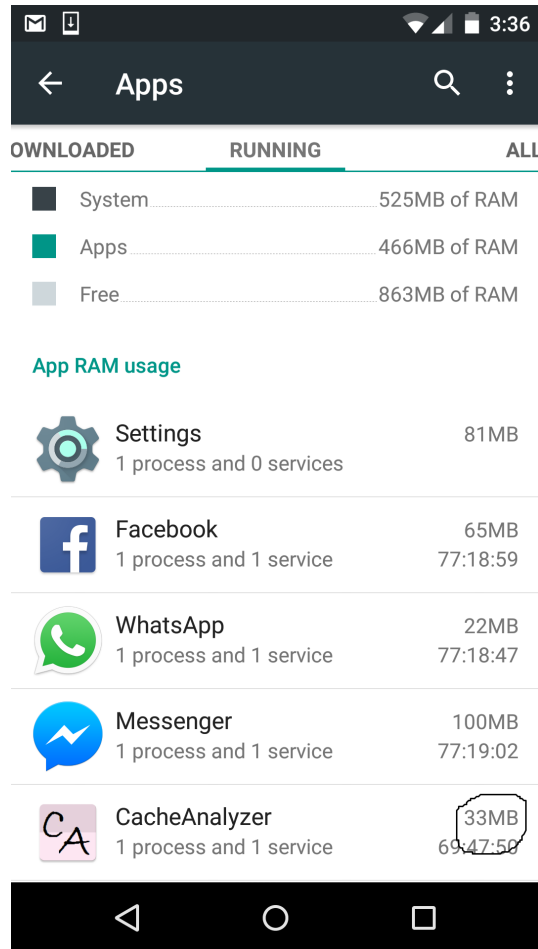


Figure 3-6: CacheAnalyzer is an application that collects the default (LRU) cache metrics. The service mentioned in the diagram is the background service that updates the metrics every second. We'll discuss the design of this application in Section 4.4 but it's worth noting here that the memory footprint of the application is quite low. This is not to suggest that it won't get killed by the OOM killer, which is why we're backing the data up.



# Chapter 4

## Experiment Setup and Application Design

### 4.1 Introduction

In the previous chapter, we examined the metrics we needed to analyze the best caching scheme for the APC. We looked into some of the common CRA(s) and analyzed some of the challenges in collecting the metrics needed to determine if contextual information can help better the CHR of the Hybrid approach. To analyze the efficiencies of the default LRU and the Hybrid approaches, two Android applications namely *CacheAnalyzer* and *ContextAnalyzer* were written and installed in the APMD(s) of research volunteers. In this chapter, we'll address the details of the experiment and analyze the design of these two Android applications.

### 4.2 Eligibility for Volunteers

The following criteria had to be met to qualify as a research volunteer:

- The volunteer must own an APMD with Android OS version greater than 5.0 (Lollipop and above).

- The volunteer must be willing to install two third party applications on their APMD for the duration of a week.

The algorithm that obtains the list of processes currently residing in memory and the algorithm that retrieves the current foreground application do not apply to Android OS versions below 5.0, hence the necessity for APMD(s) with OS versions that are Lollipop and above.

#### 4.2.1 Addressing Privacy Concerns

Since the research volunteers had to install two third party applications (*CacheAnalyzer* and *ContextAnalyzer*), it was crucial to address any privacy concerns that arose. In order to gain the trust of the volunteers, we adopted a policy of full transparency and took the following measures:

- We fully explained to each volunteer, the nature of the data we were collecting i.e. cache metrics.
- We open sourced both Android applications [50][51] in case the volunteers wanted to verify that there were no nefarious activities going on in the background.
- We did not automatically store the cache metrics in a server. Instead, we asked the volunteers to send us the data in order to convey the message that we weren't collecting any private data that the volunteers weren't aware of.

Although the aforementioned points address the privacy issues, we realize that divulging the nature of the experiment has the potential to bias the volunteers in their APMD usage. We explore this aspect in detail in the Points of Weaknesses section.

### 4.3 Process and Duration of Experiment

### 4.4 Cache Analyzer Design

### 4.5 Context Analyzer Design

### 4.6 Tools, Version Control and APK

#### 4.6.1 Android Studio

#### 4.6.2 Git

#### 4.6.3 Third Party APK(s)



# Chapter 5

## Data Analysis and Results

### 5.1 Introduction

### 5.2 Phase A Data

### 5.3 Phase B Data

### 5.4 Default Approach Metrics

### 5.5 Pure Prediction Approach Metrics

### 5.6 Hybrid Approach Metrics

### 5.7 Factors Influencing Variation In Data

#### 5.7.1 Android Version and Phone Model

#### 5.7.2 Number of Applications Installed

#### 5.7.3 Number of Applications Used

#### 5.7.4 User Bias

#### 5.7.5 User Demographics



# Chapter 6

## Scope for Future Work

### 6.1 Introduction

### 6.2 Improving Context Analysis

#### 6.2.1 Machine Learning in Calendar Parsing

#### 6.2.2 Calendar Parser - LFU Hybrid

#### 6.2.3 Other Ways of Gathering User Behavior Data

#### 6.2.4 Custom Priorities in Hybrid Cache

### 6.3 Alternate Ways of Using the Contextual Information

#### 6.3.1 Switching to Silent Mode

#### 6.3.2 Disabling Texting at High Speeds

### 6.4 Implementing the Hybrid Cache

## Chapter 7

## Conclusion



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