

# 基于统计分析的自适应 Android系统性能异常 检测

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# Background

# Android

- Most popular smart phone operating system (compete with iOS)
- Open source, which cause fragmentation.
- Third-party modifications, devices with different hardware, different version exist in market.
- Performance degradation upon use with time lapse.
- Performance anomaly occurs frequently which need analysis.
- To solve performance issue, first find and understand anomaly.



# Performance anomaly

- Anomaly — different from normal
  - High latency
  - UI unresponsiveness
  - Disk IO
  - Memory swapness
  - Context switch
- Anomaly may not be too obvious to easy detected by human.
- May only exist in subtle statistics.

Related work

# Related work

- Detecting large-scale system problems by mining console logs
  - Finding problem from log in large scale datacenter which is hard for manually inspection.
- Fingerprinting the Datacenter: Automated Classification of Performance Crises
  - Matching runtime statistics fingerprint with existing database, detecting similar problem and match possible solutions.
- AppInsight: Mobile App Performance Monitoring in the Wild
  - Inspecting calling relationship and execution path by injecting code to existing app.
- Measuring the Performance of Interactive Applications with Listener Latency Profiling
  - Injecting to callback listener record logs for Java GUI application

# Methodology

# Statistics from Linux kernel

- Android is based on Linux kernel
  - /proc filesystem
  - cpuinfo
  - meminfo
  - process
  - io
  - network
- 
- We can have enormous runtime statistics either for the whole system or for a single process, and no need for modification on application.



# Proc filesystem

```
$ ls
1      8700      config.gz  execdomains  irq         kpageflags  mtrr       slabinfo     timer_list
22576  8726      consoles  fb           kallsyms    loadavg     net        softirqs     timer_stats
22602  8727      cpuinfo   filesystems  kcore       locks       pagetypeinfo  stat         tty
22603  acpi      crypto    fs           keys        mdstat      partitions  swaps        uptime
22817  buddyinfo  devices  interrupts  key-users   meminfo     sched_debug  sys          version
2663   bus      diskstats iomem        kmsg        misc        schedstat    sysrq-trigger  vmallocinfo
69     cgroups   dma       ioports      kpagecgroup modules      scsi        sysvipc      vmstat
783    cmdline  driver    ipmi         kpagecount  mounts      self        thread-self   zoneinfo
```

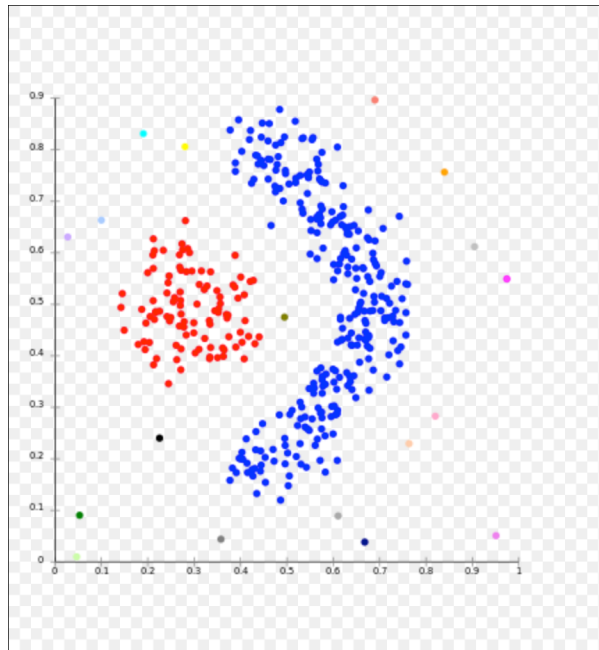
```
$ ls
attr      cmdline  environ  limits  mounts  oom_adj  root  stat  wchan
autogroup  comm     exe      map_files  mountstats  oom_score  sched  statm
auxv       coredump_filter  fd       maps      net        oom_score_adj  schedstat  status
cgroup     cpuset   fdinfo   mem       ns         pagemap      smaps    syscall
clear_refs  cwd      io       mountinfo numa_maps  personality  stack    task
```

# Performance anomaly detection

- When does a computer system run into an anomaly state.
- Large amount of statistics makes it hard for human observation.
- What is anomaly? High load? High latency? Hardware exception?
- Anomaly is rare in real system, difficult for direct prediction.
- Unsupervised learning from normal data — anomaly detection.
- Find pattern adaptively from real time usage data without human intervention.

# Anomaly(outlier) detection

- Gather normal feature pattern to form cluster.
- Outlier is predicted to be those far from any cluster.



# UI latency prediction

- Find when a Android app encounter high UI latency with high-speed photography.
- Runtime statistics from Android OS kernel and framework, both by Linux kernel and code injection to framework and application.
- Can get labeled data about when we encounter UI latency and the runtime state.
- Enable further analysis of its reason.
- Maybe build a supervised machine learning model to predict latency directly from statistics.

# Schedules

# Schedules

- 2017.1 – 2017.3 build a tool which can gather and store runtime statistics from Linux kernel in proper format.
- 2017.3 – 2017.4 gather runtime statistics (both normal and anomaly) in experiment settings.
- 2017.4 – 2017.5 train unsupervised model for anomaly detection and online testing.
- 2017.5 – 2017.6 synthesize and thesis writing.

Thanks