
Toward Practical Automation for Software Engineering

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Assistant Professor

CS Research Day

George Mason University

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Software Engineering

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The methods and techniques by which developers design, create, test, and manage software

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My research goal: Design tailored *automated approaches* to help facilitate *developer needs* throughout the software development and maintenance lifecycle.

PRACTICAL SIGNIFICANCE

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Blend *scientific discovery* with *practical significance*

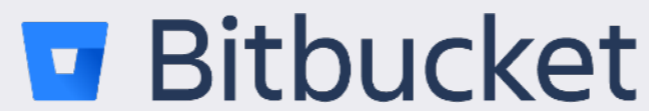
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How Can We Design Practical Automation?

MINING SOFTWARE REPOSITORIES



MINING SOFTWARE REPOSITORIES

MINING SOFTWARE REPOSITORIES



Source Code
Files



Software
Documentation



Screenshots



Screen
Recordings

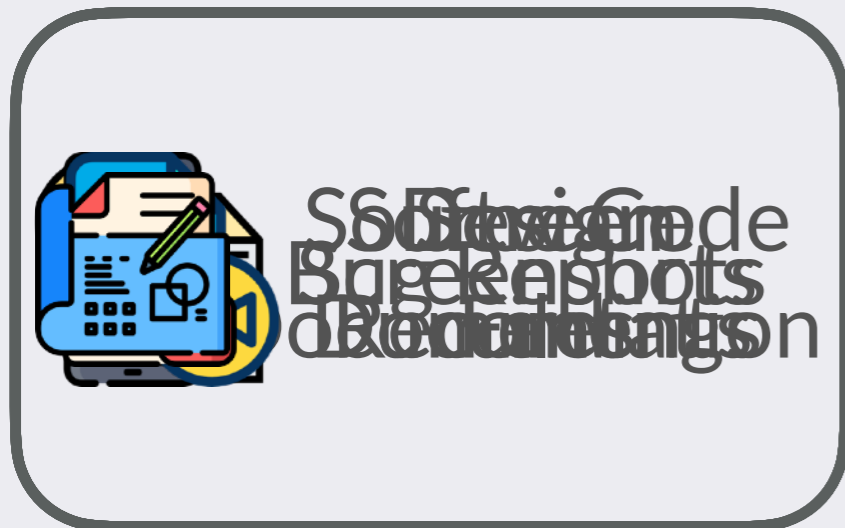


Bug Reports



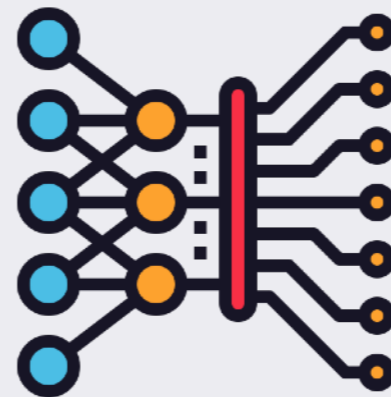
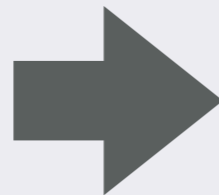
Design
Documents

LEARNING PATTERNS FROM SOFTWARE DATA

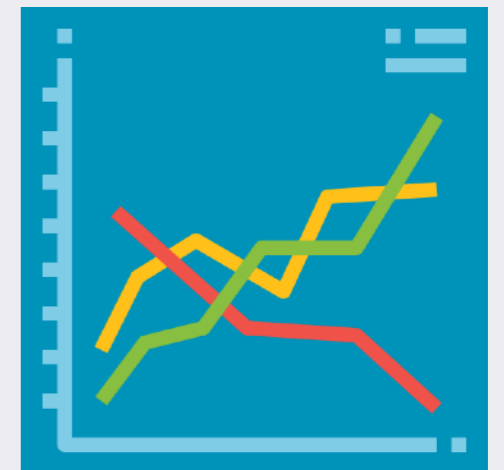
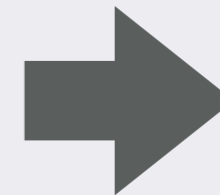


LEARNING PATTERNS FROM SOFTWARE DATA

Software
Repository Data

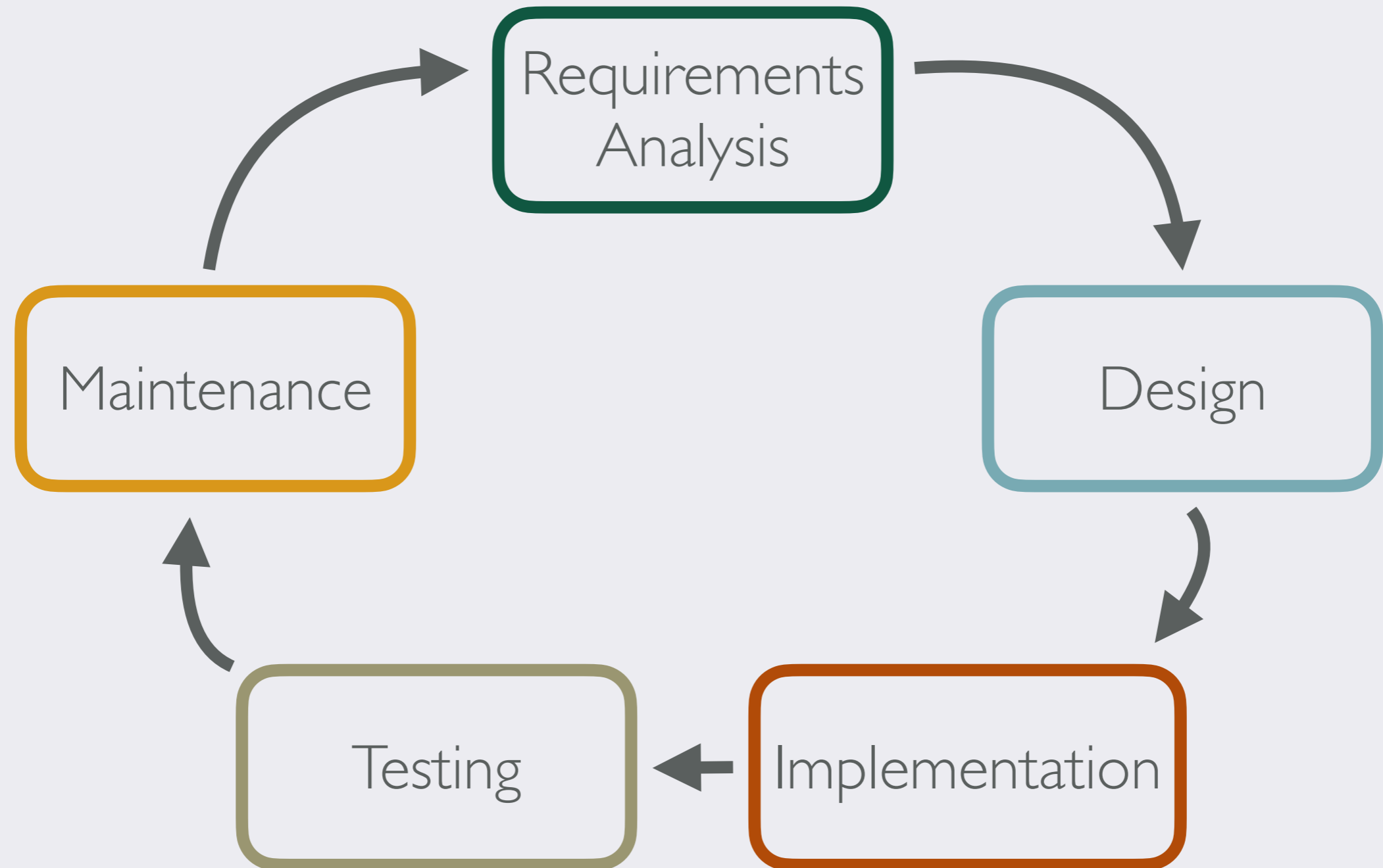


Machine
Learning



Salient
Patterns

SOFTWARE DEVELOPMENT LIFECYCLE



OVERVIEW

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1. A Brief Look into my Research

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- Automatically Prototyping Mobile App GUIs

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- Automatically Prototyping Mobile App GUIs
- Translating Screen Recordings into Replayable Scenarios

OVERVIEW

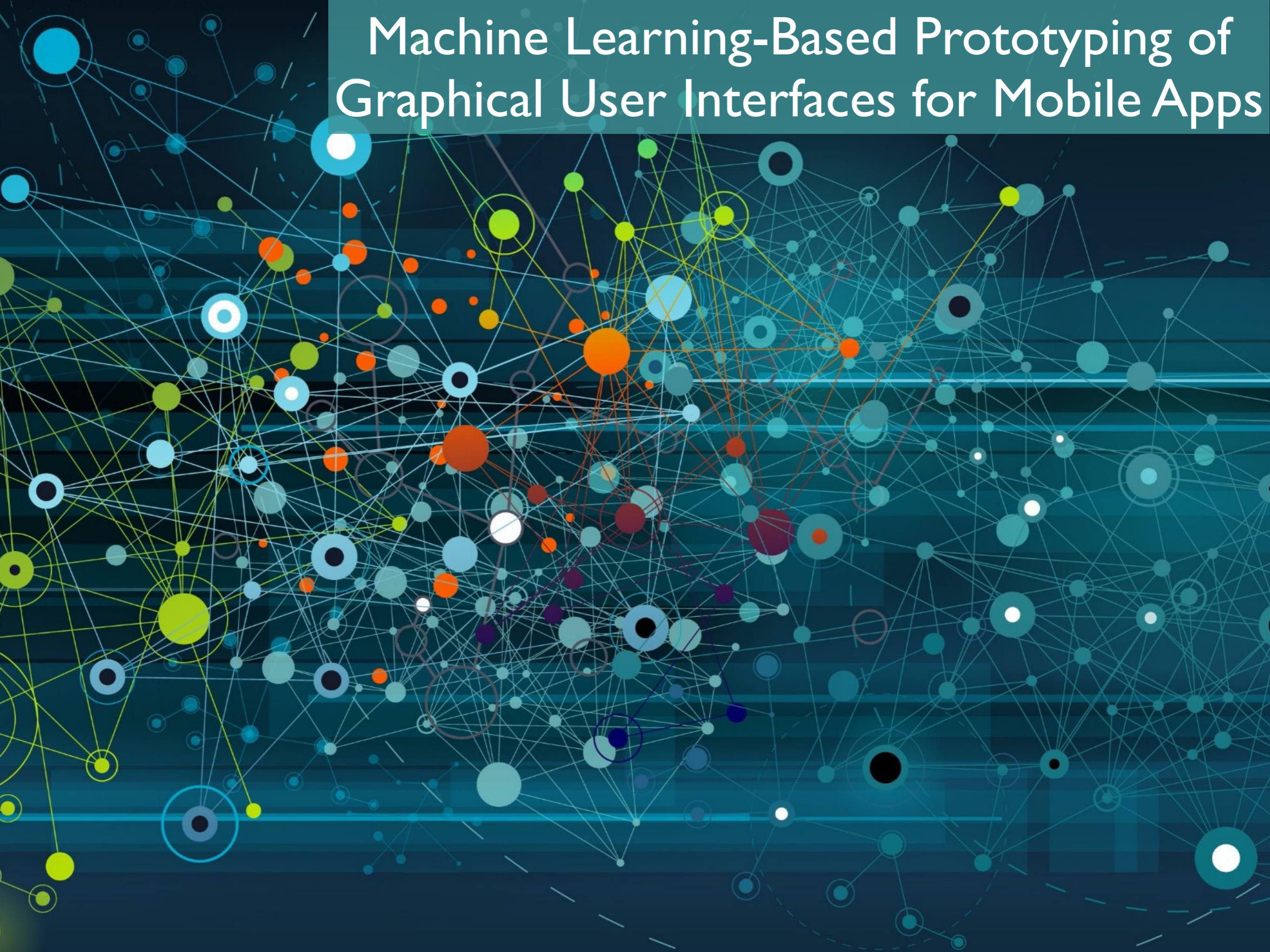
1. A Brief Look into my Research

- Automatically Prototyping Mobile App GUIs
- Translating Screen Recordings into Replayable Scenarios

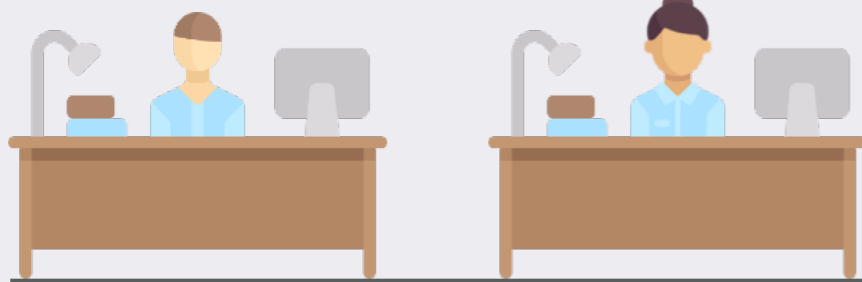
2. Overview of other Research Areas & Future Work

A BRIEF LOOK INTO MY RESEARCH

Machine Learning-Based Prototyping of Graphical User Interfaces for Mobile Apps



RESEARCH PROBLEM



UI/UX Design Team



Development Team

RESEARCH PROBLEM

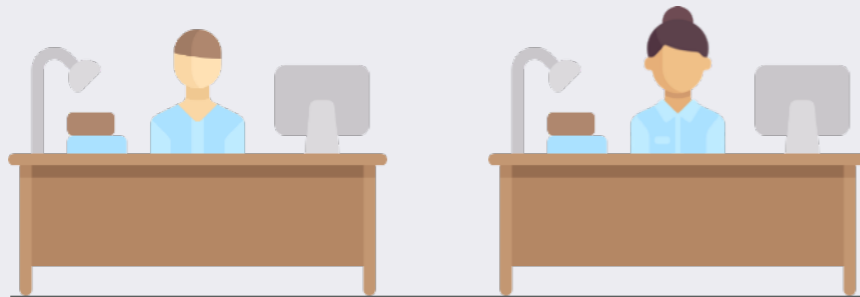
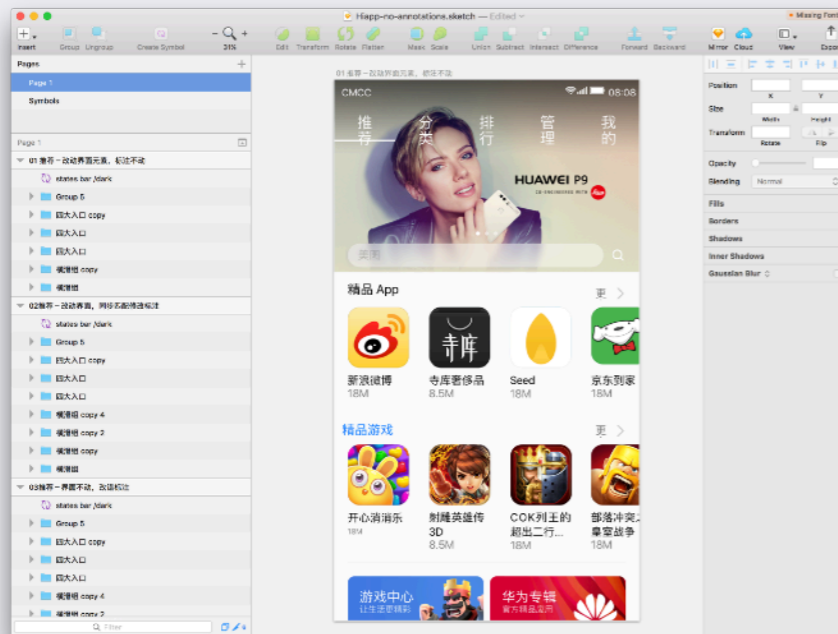


UI/UX Design Team



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RESEARCH PROBLEM



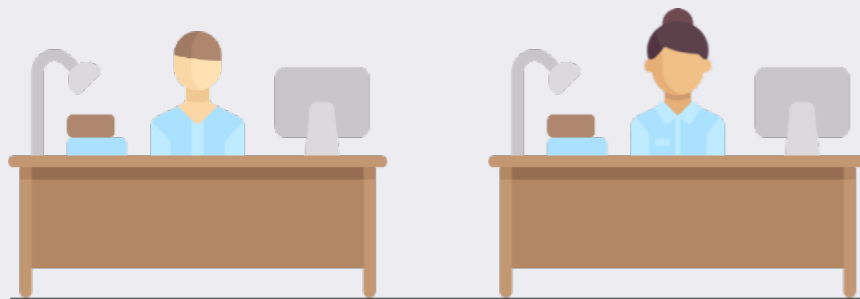
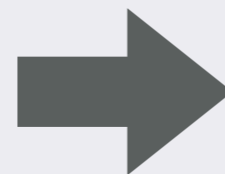
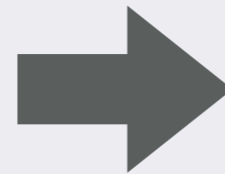
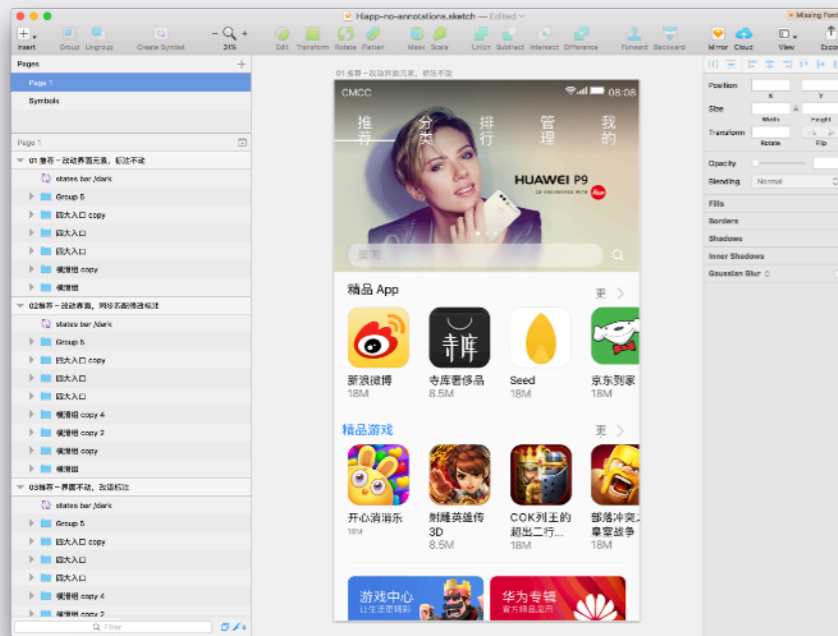
UI/UX Design Team



Development Team

RESEARCH PROBLEM

Prototype GUI Code

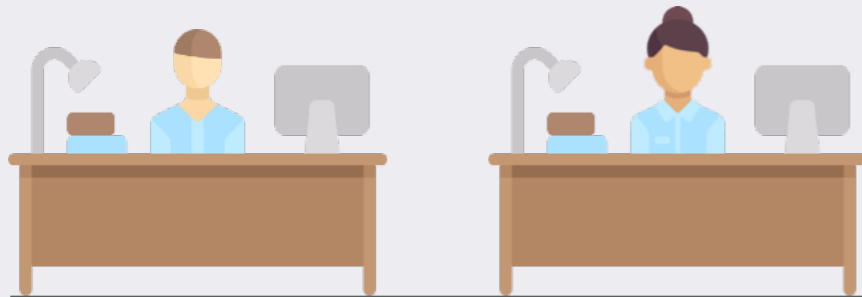
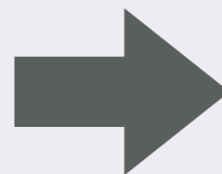
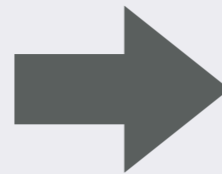
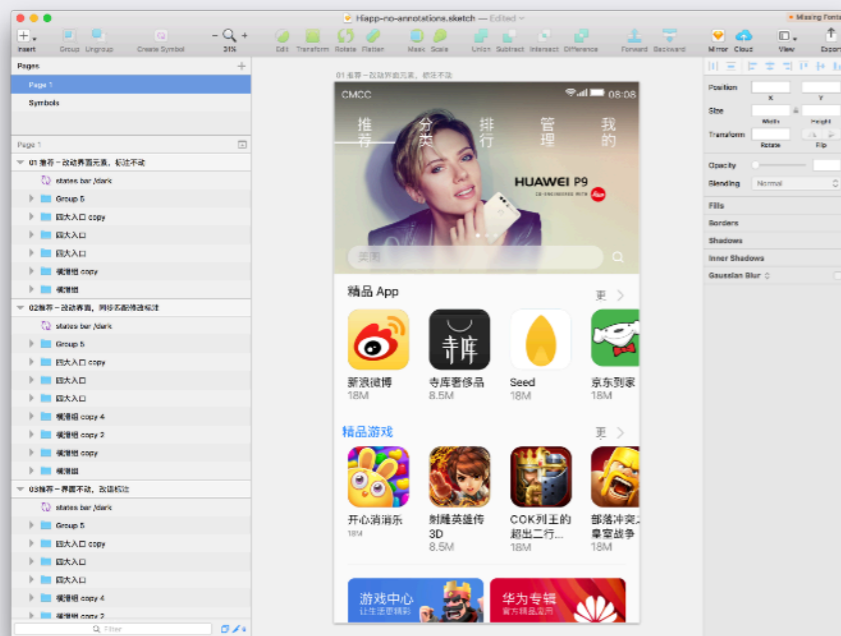


UI/UX Design Team

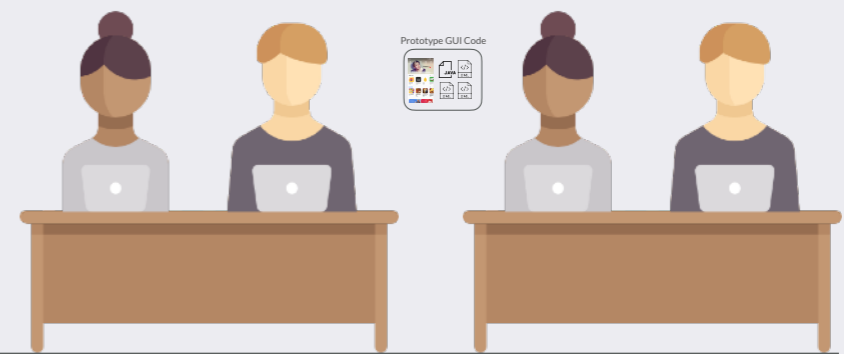


Development Team

RESEARCH PROBLEM



UI/UX Design Team



Development Team

FIRST PRINCIPLES

How does a developer translate a GUI mock-up into code?

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1. ***Detect*** or identify GUI-components that exist in a mock-up
2. ***Classify*** these GUI-components into their constituent types
3. ***Assemble*** the GUI-components into a hierarchy and stipulate styles

FIRST PRINCIPLES

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Detection

FIRST PRINCIPLES

Detection Classification

FIRST PRINCIPLES

Detection

Classification

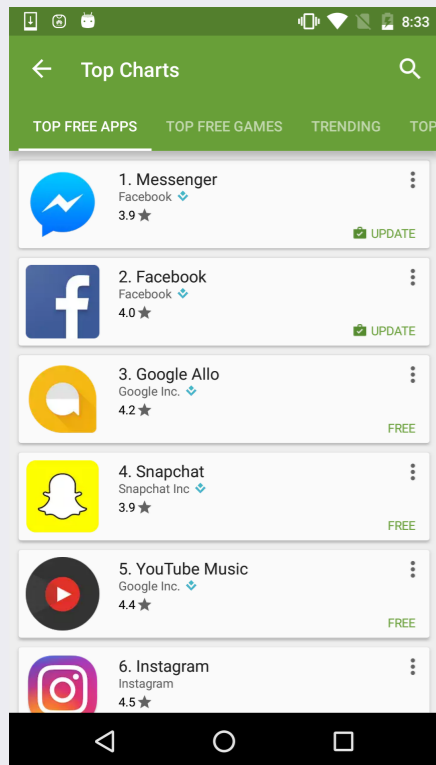
Assembly

FIRST PRINCIPLES

Detection — Classification — Assembly

PHASE I: GUI-COMPONENT DETECTION

Computer Vision-Based Detection*

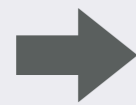
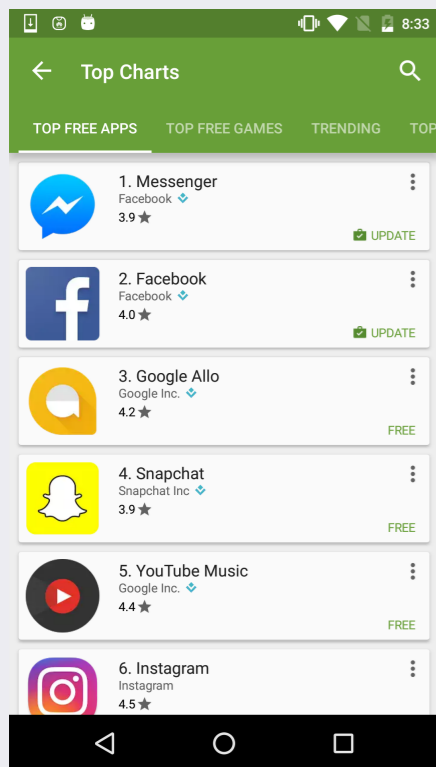


Input Image

*Tuan Anh Nguyen and Christoph Csallner. 2015. Reverse Engineering Mobile Application User Interfaces with REMAUI, In Proceedings of the 2015 30th IEEE/ACM International Conference on Automated Software Engineering (ASE '15)

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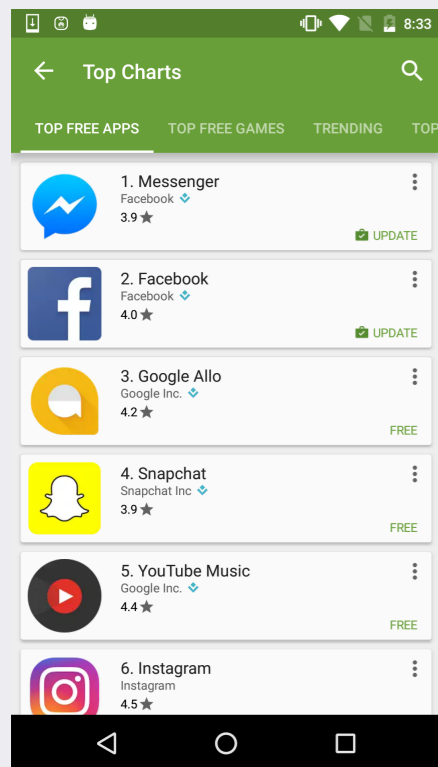


Input Image

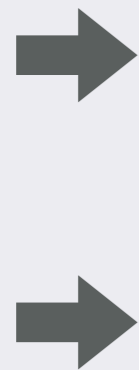
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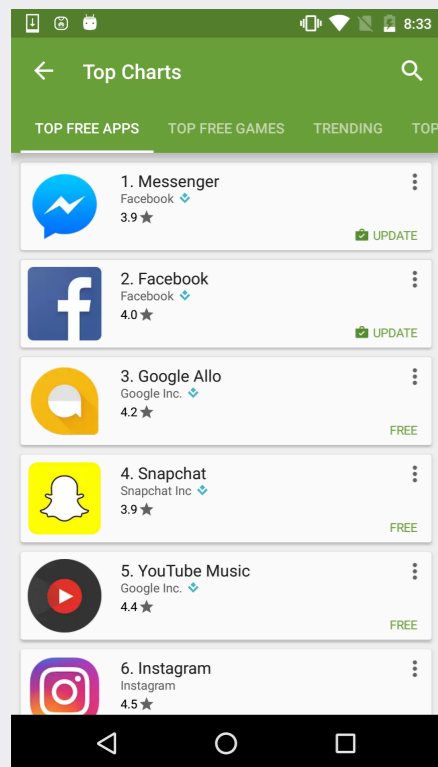


Canny's Edge
Detection

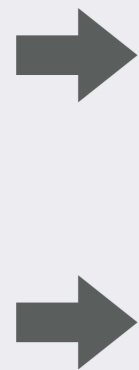
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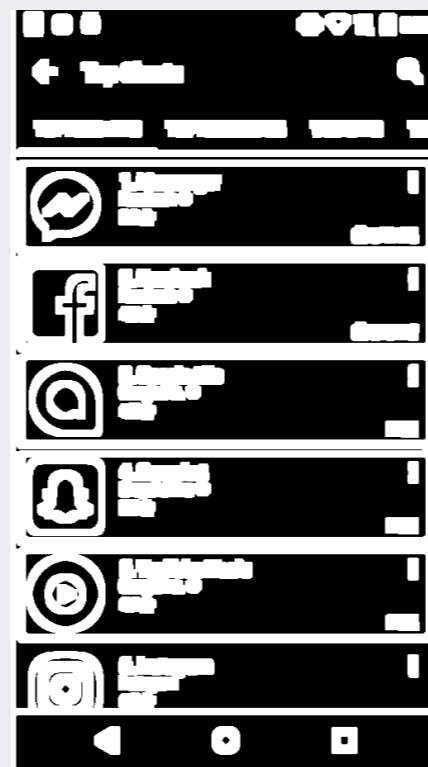
Computer Vision-Based Detection*



Input Image



Canny's Edge
Detection

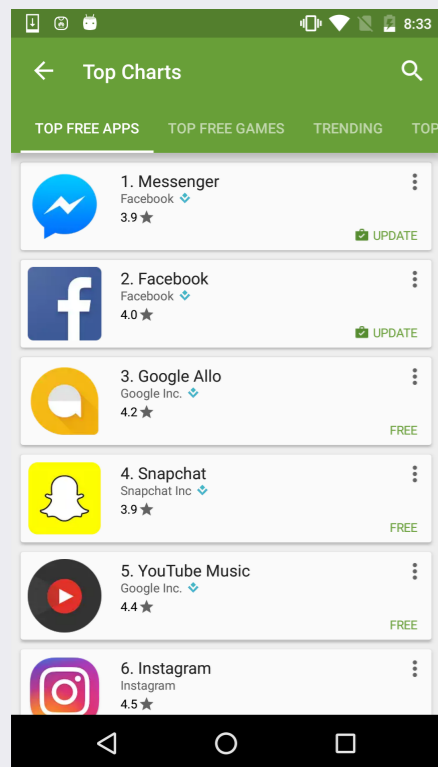


Edge
Dilation

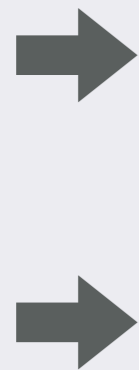
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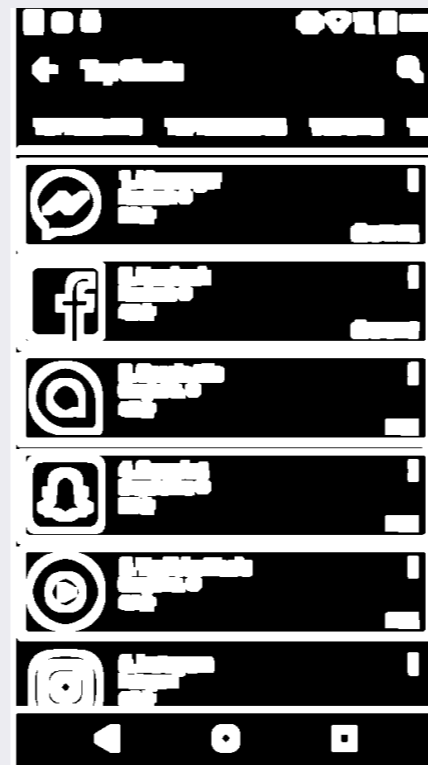
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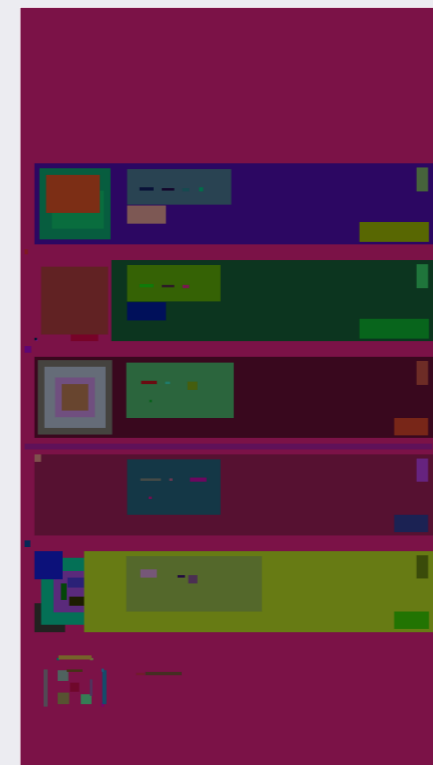
Input Image



Canny's Edge
Detection



Edge
Dilation

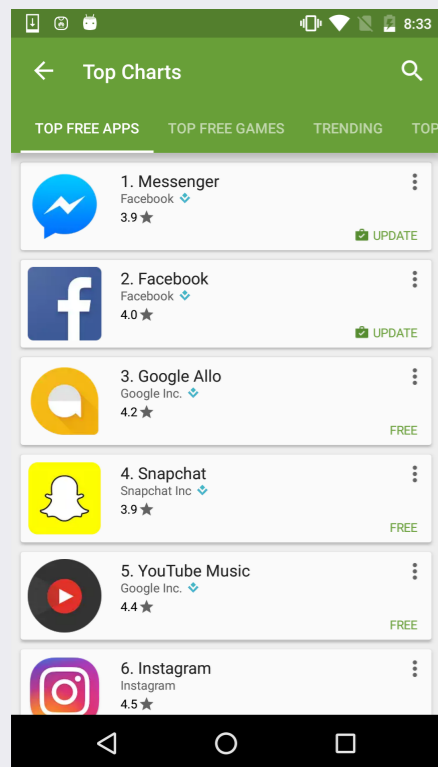


Contour
Bounding Boxes

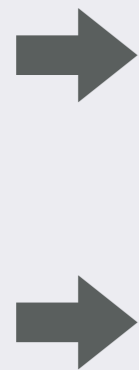
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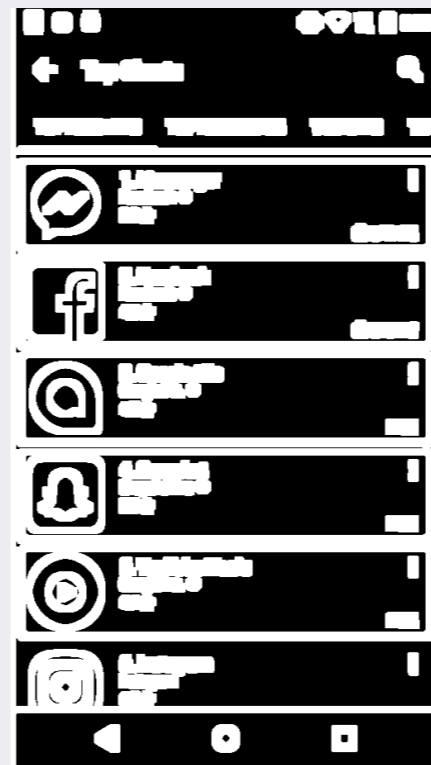
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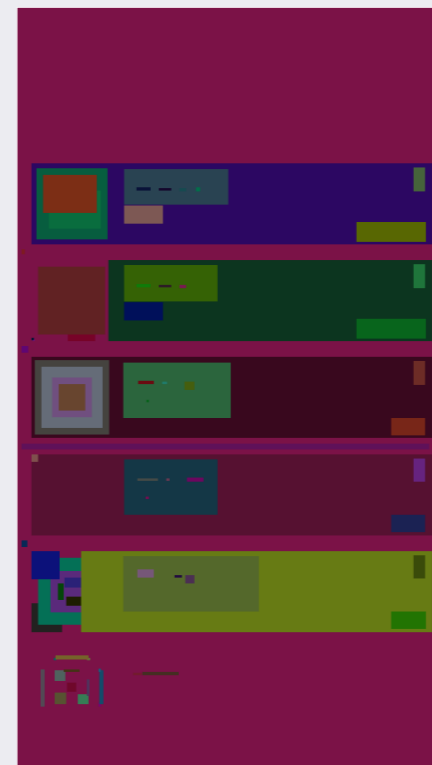
Input Image



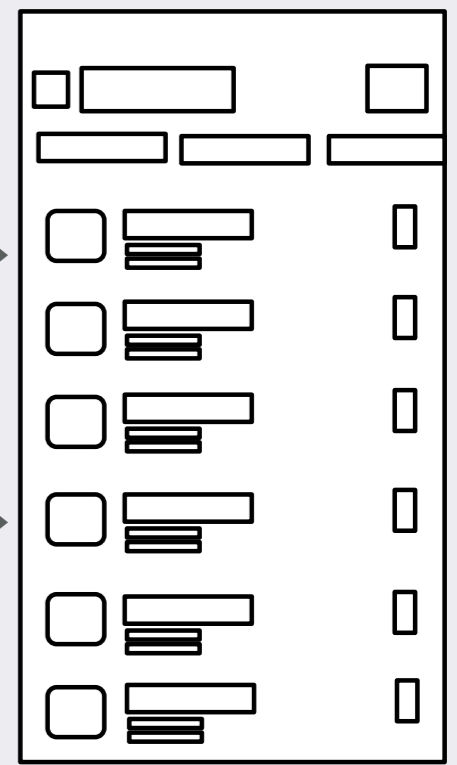
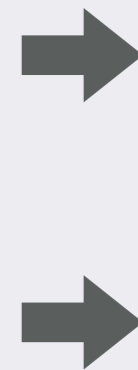
Canny's Edge
Detection



Edge
Dilation



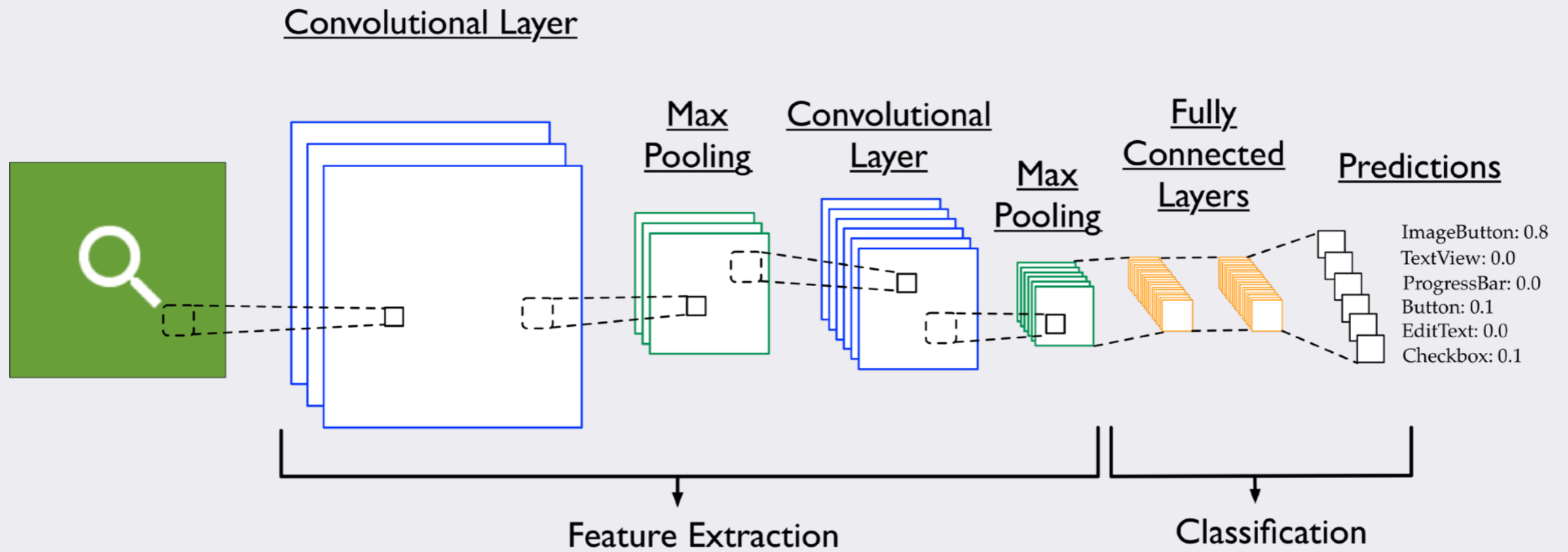
Contour
Bounding Boxes



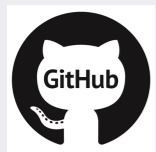
GUI-Component
Bounding Boxes

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PHASE 2: GUI-COMPONENT CLASSIFICATION



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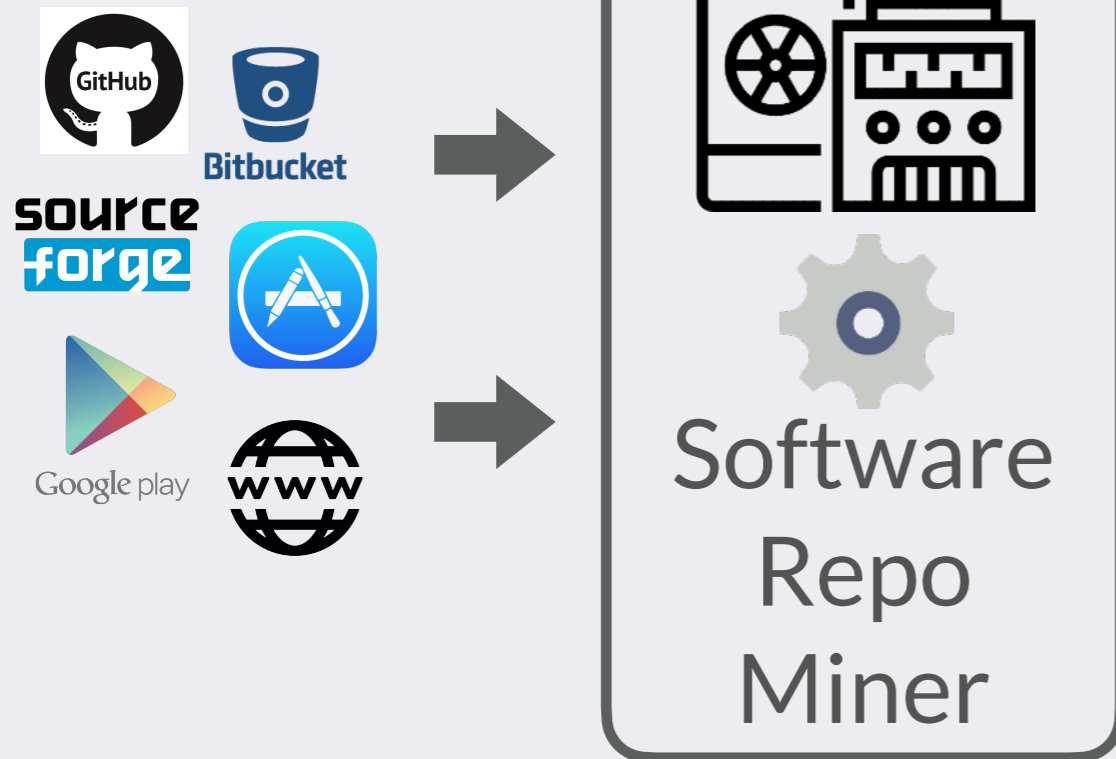
Bitbucket



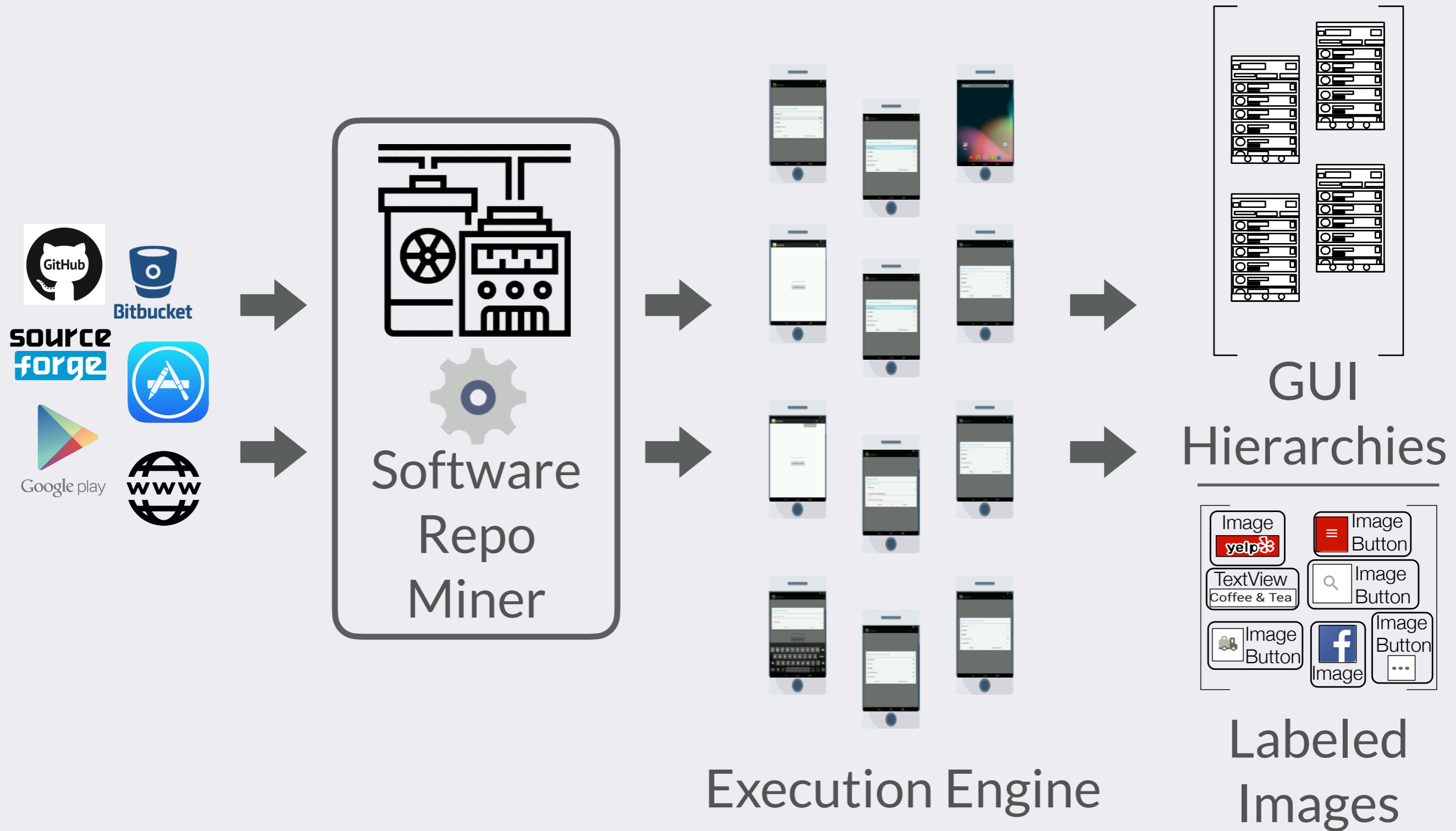
Google play



PHASE 2: GUI-COMPONENT CLASSIFICATION

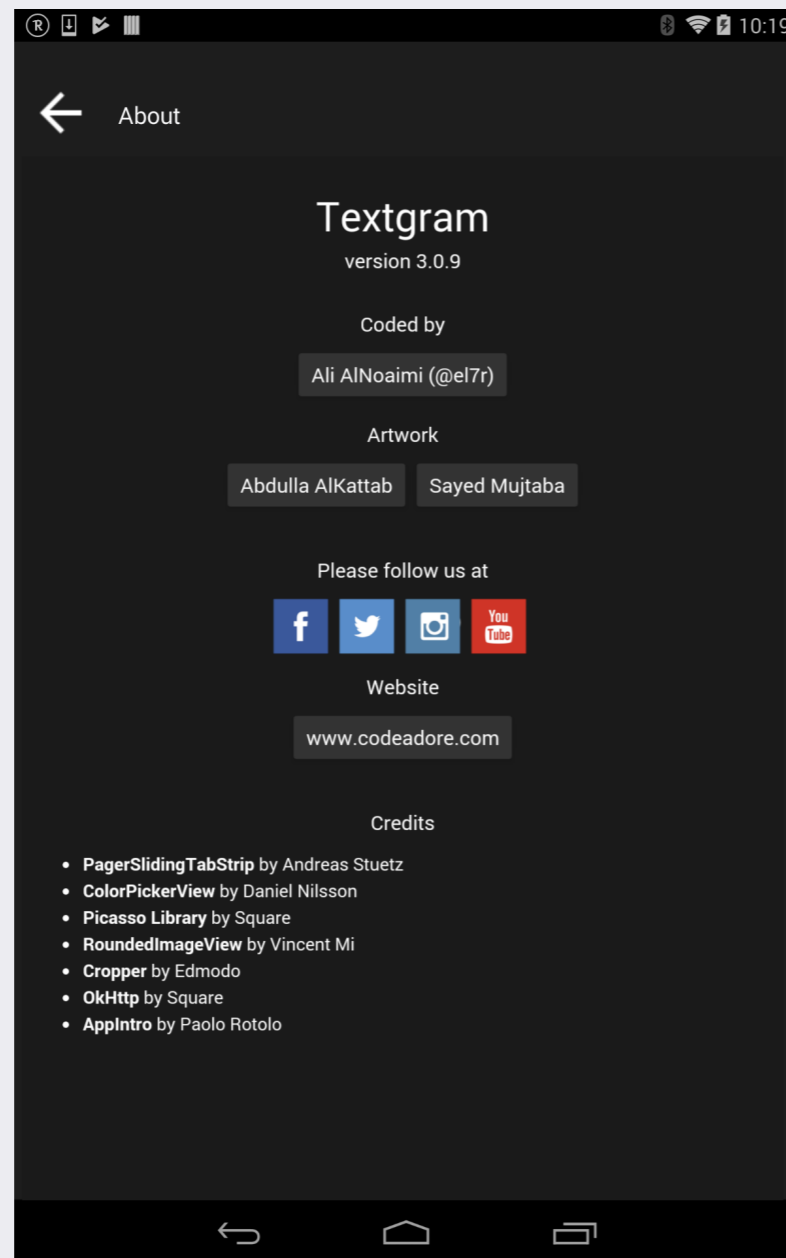


PHASE 2: GUI-COMPONENT CLASSIFICATION

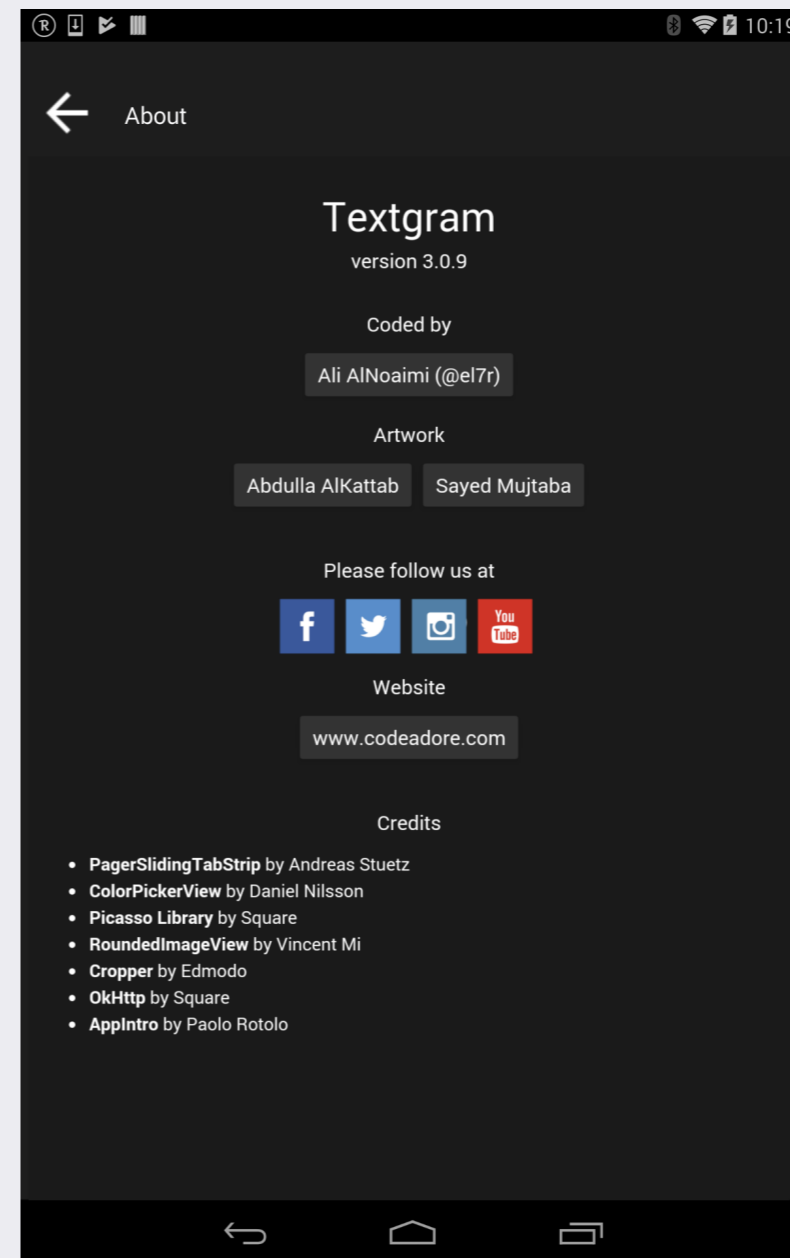


APPLICATIONS GENERATED BY REDRAW

Textgram



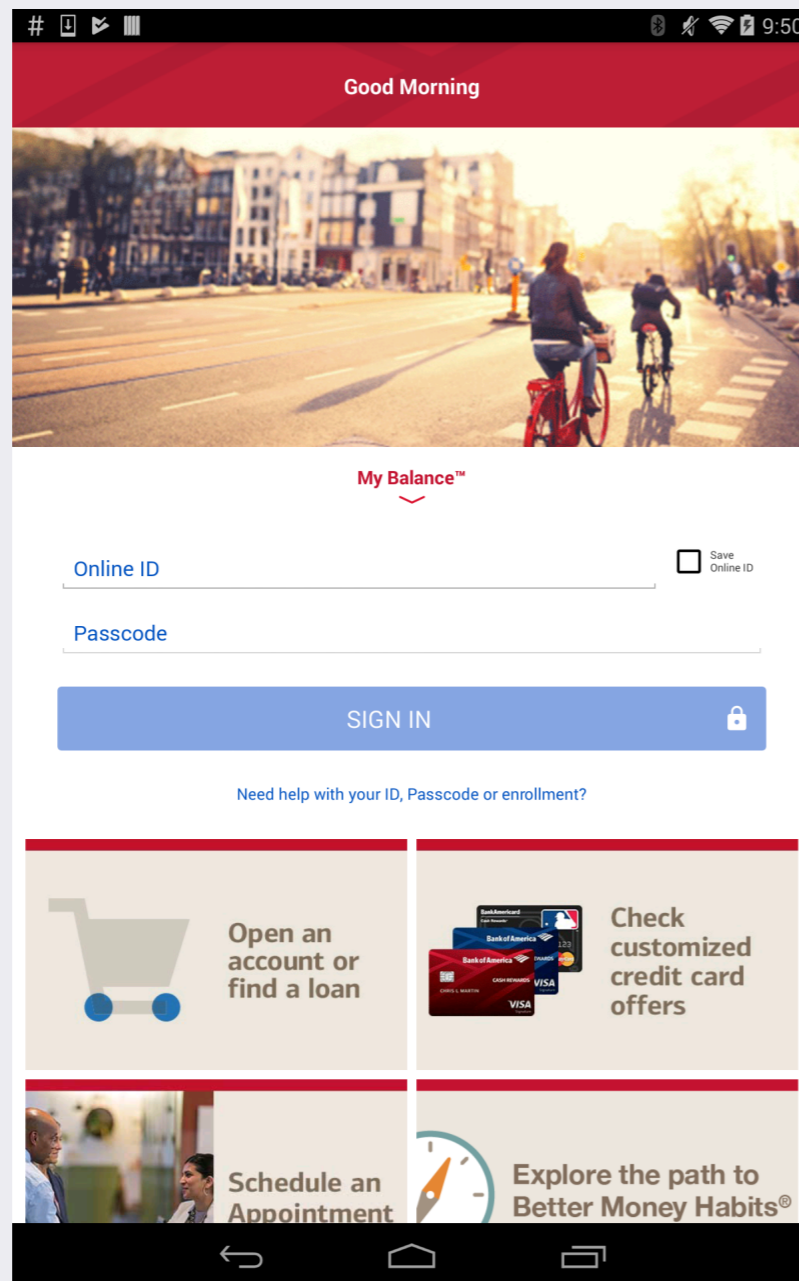
A) Original Application



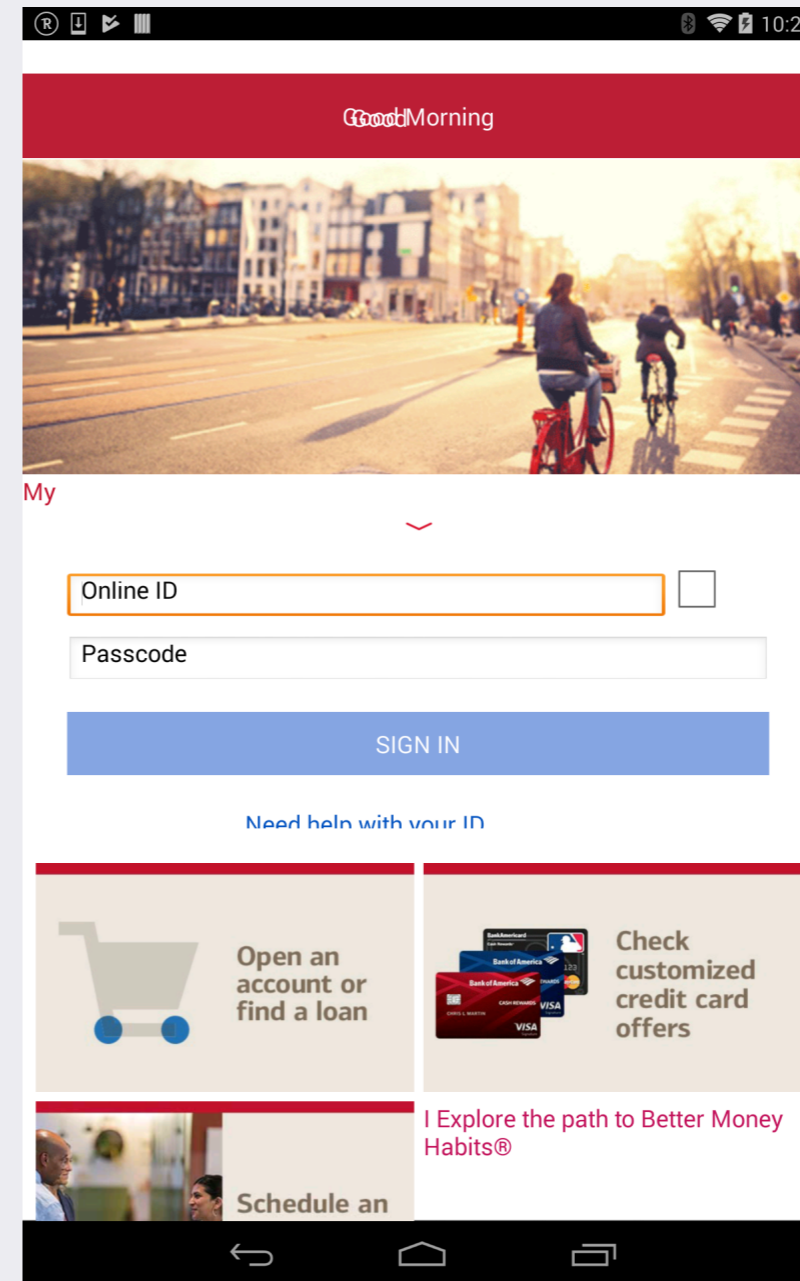
B) ReDraw App (MockUp)

APPLICATIONS GENERATED BY REDRAW

Bank of America



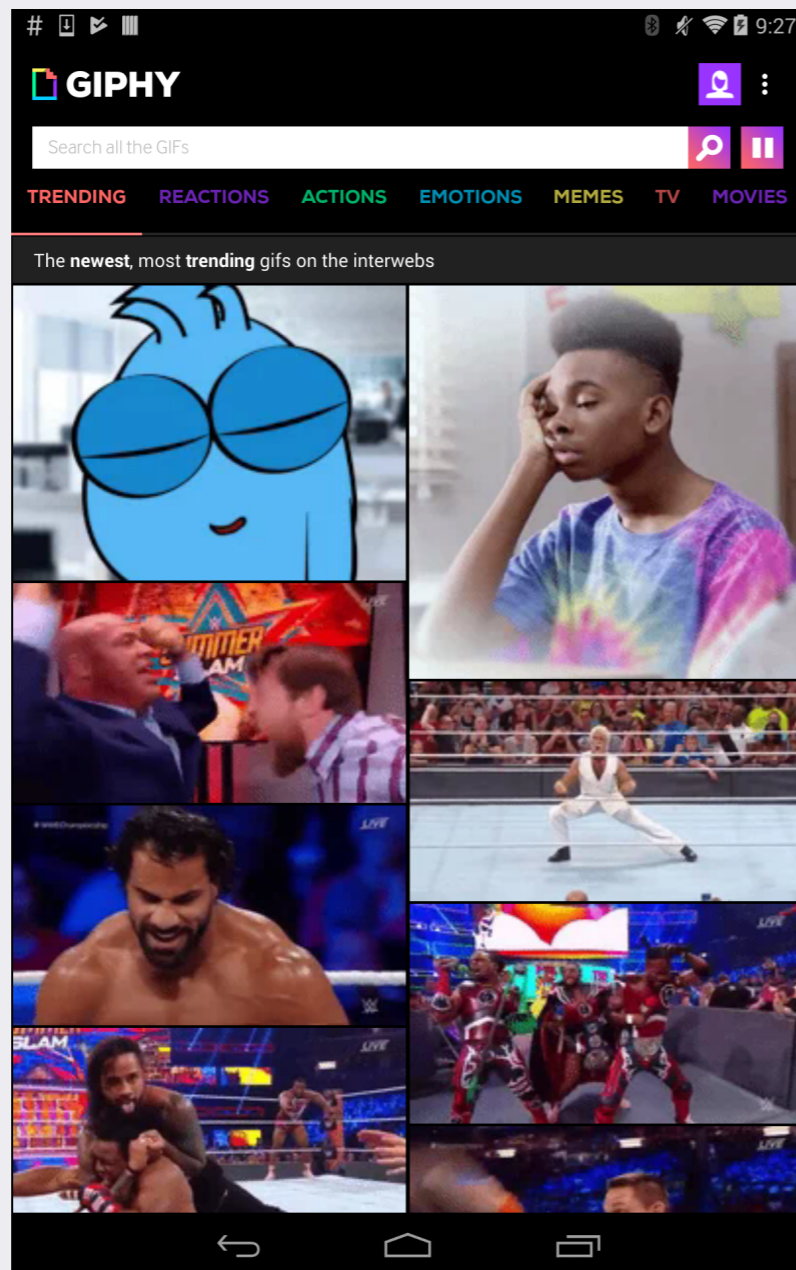
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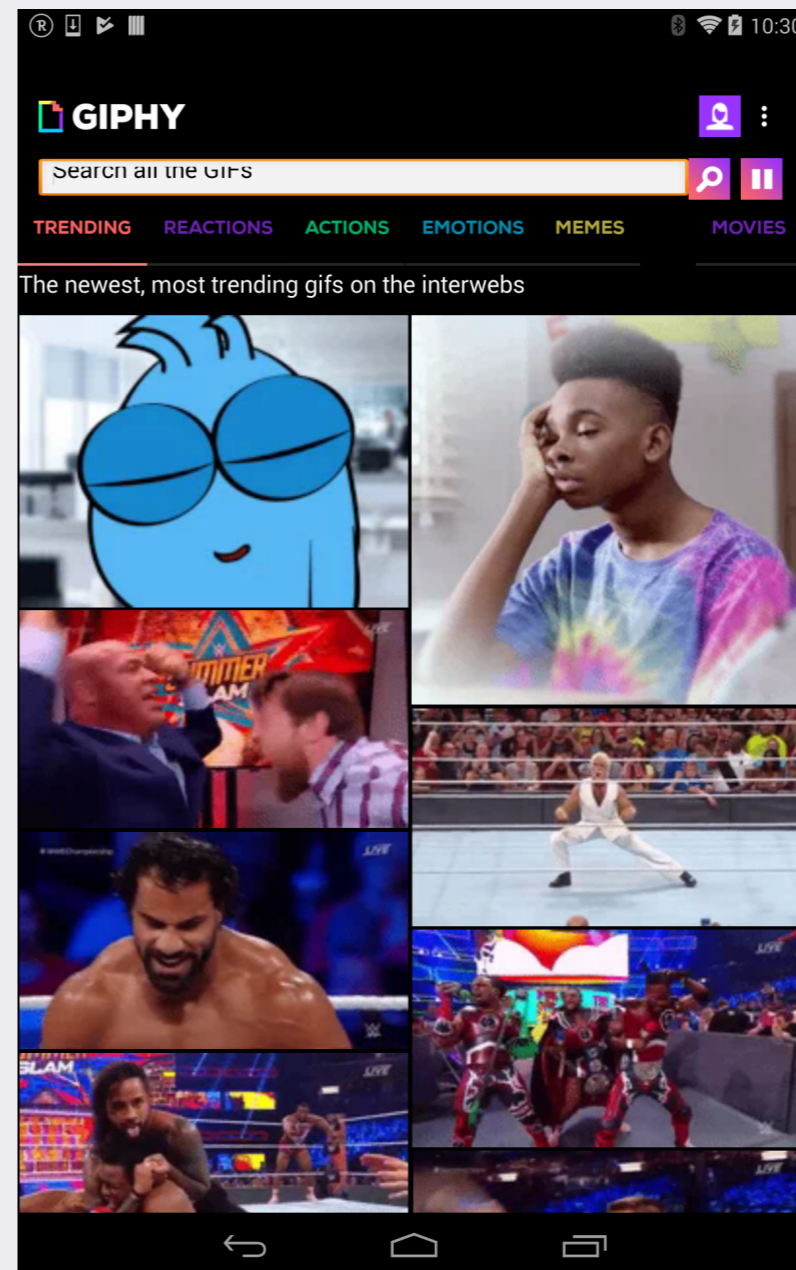
B) ReDraw App (MockUp)

APPLICATIONS GENERATED BY REDRAW

Giphy



A) Original Application

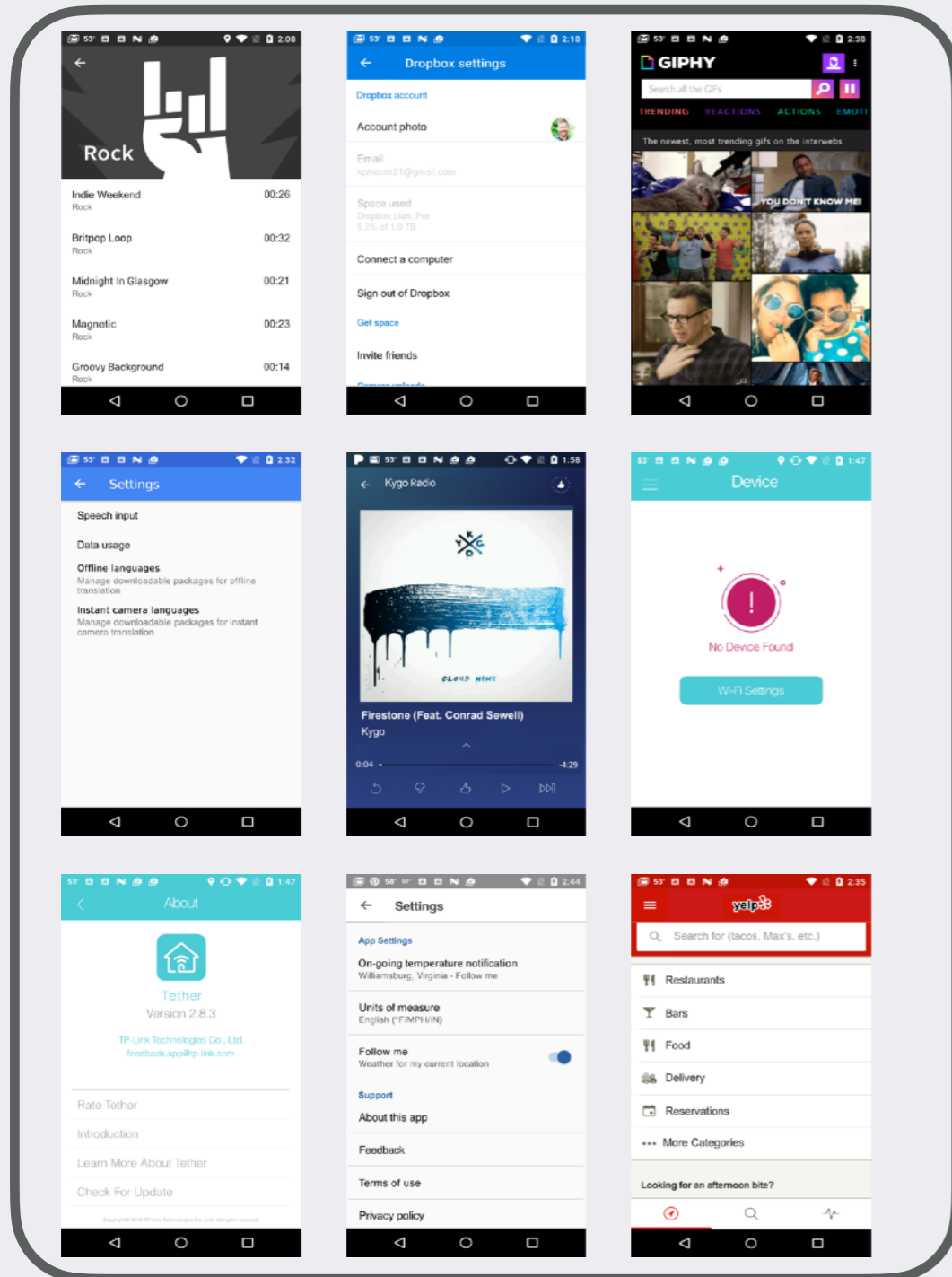


B) ReDraw App (CV)

STUDY: DEVELOPER UTILITY

9 Screens from 8 Popular Apps

Context



Mobile Front-End Developer


Mobile Designer
 **HUAWEI**

Mobile Researcher


STUDY: RESULTS

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“It’s a good starting point... From a development standpoint, the thing I would appreciate most is getting a lot of the boilerplate code done [automatically]”

STUDY: RESULTS

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“There are going to be edge cases for different layouts, but these are easily fixed after the fact”

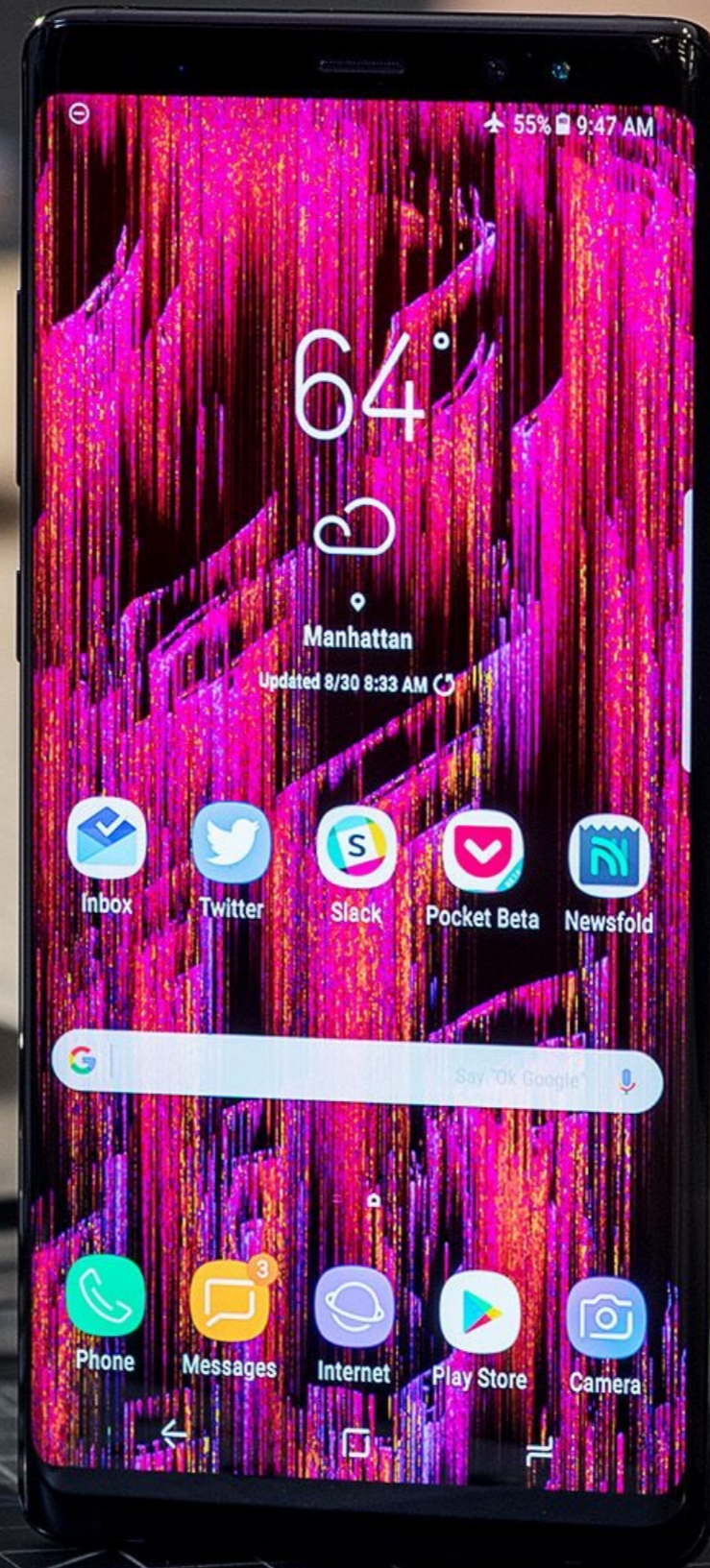
STUDY: RESULTS

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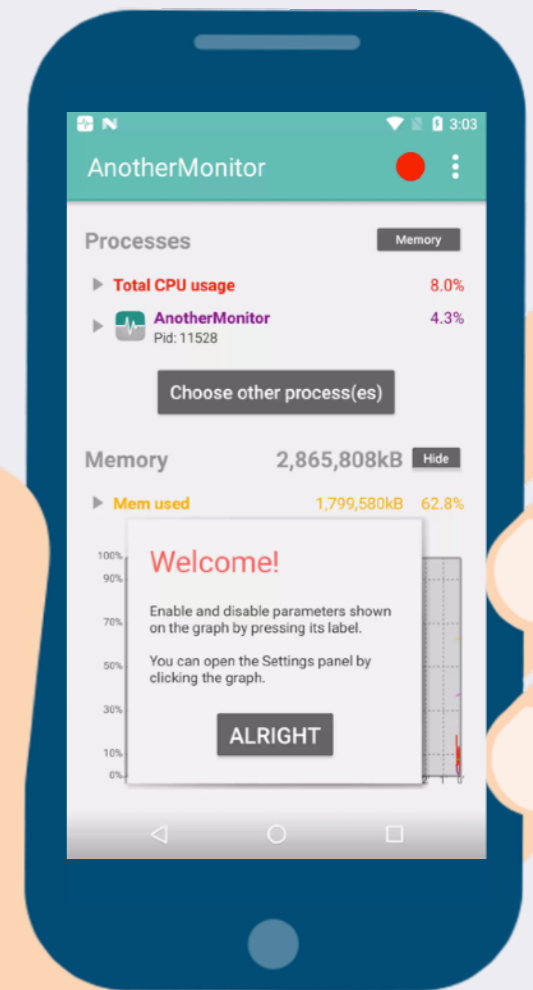
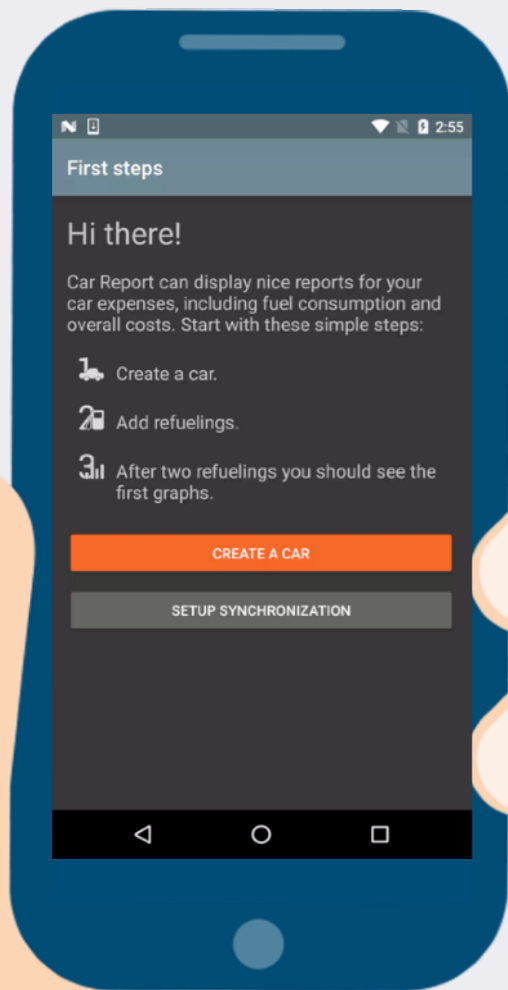
“The key thing is fast iteration. A developer could generate the initial view [using ReDraw], clean up the layouts, and have a working app. If a designer could upload a screenshot, and without any other intervention [ReDraw] could update the [existing] xml this would be ideal.”

Translating Video Recordings of Mobile Apps into Replayable Scenarios

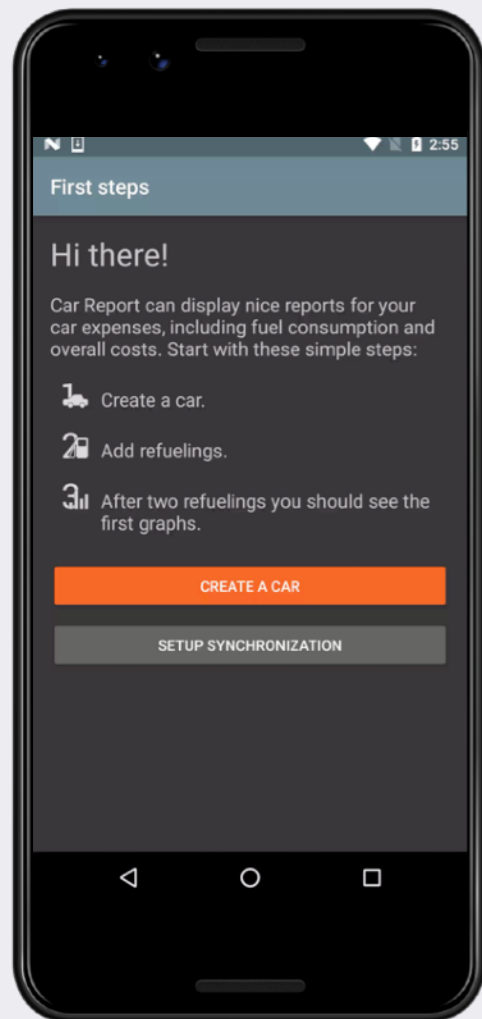


VIDEO-BASED BUG REPORTING

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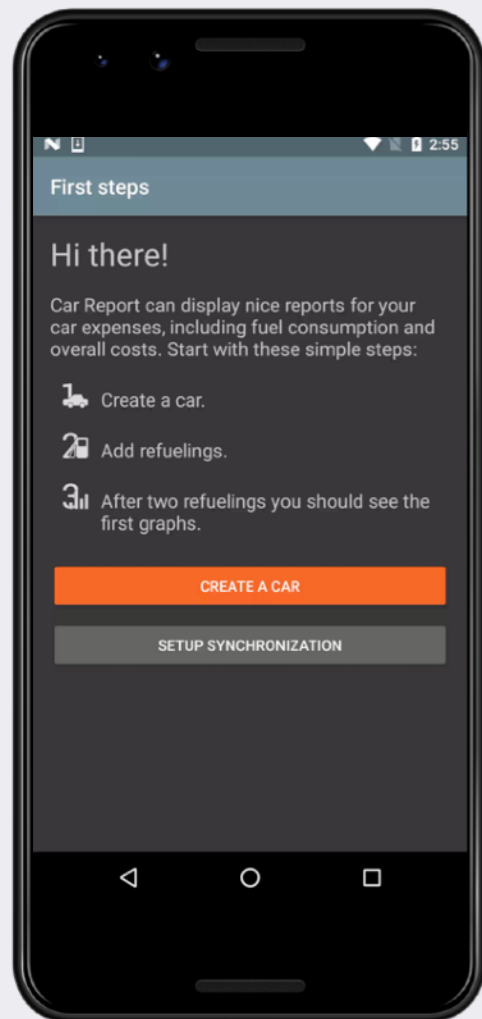


OUR SOLUTION: VIDEO2SCENARIO (V2S)



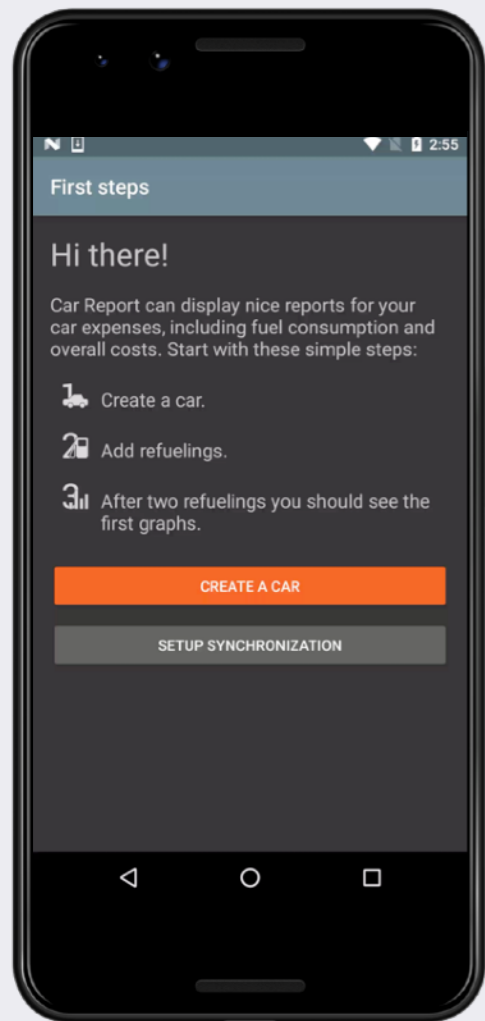
Input Screen
Recording

OUR SOLUTION: VIDEO2SCENARIO (V2S)



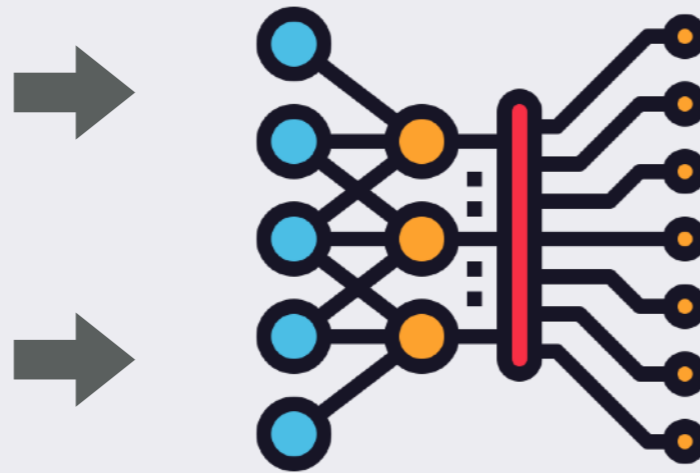
Input Screen
Recording

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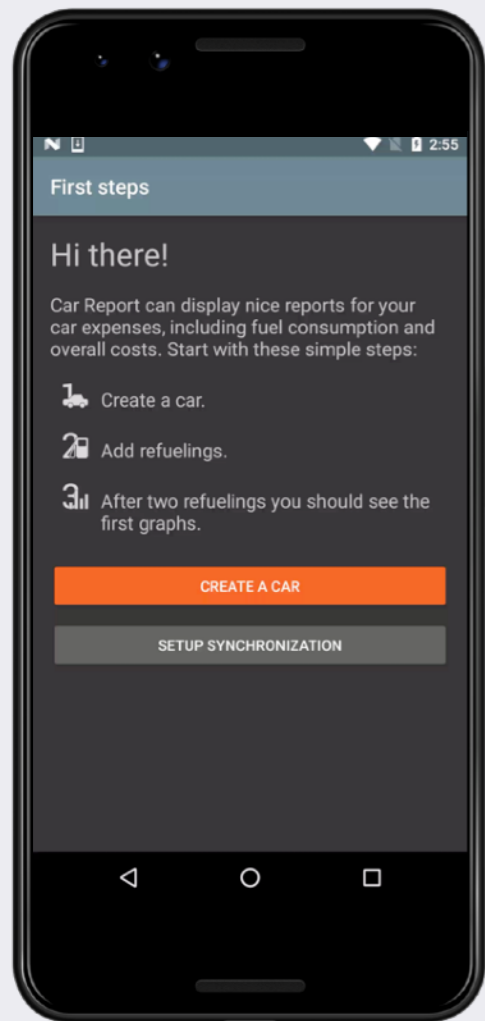


Input Screen
Recording

Neural Object
Detection &
Classification

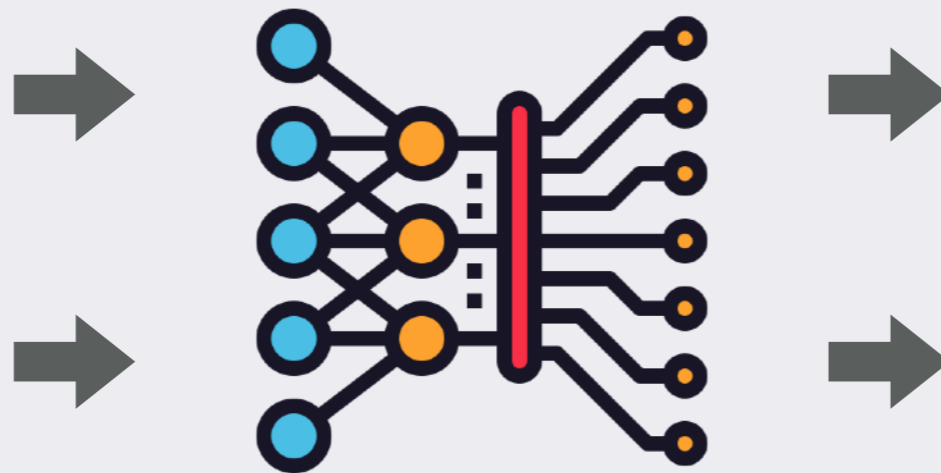


OUR SOLUTION: VIDEO2SCENARIO (V2S)



Input Screen Recording

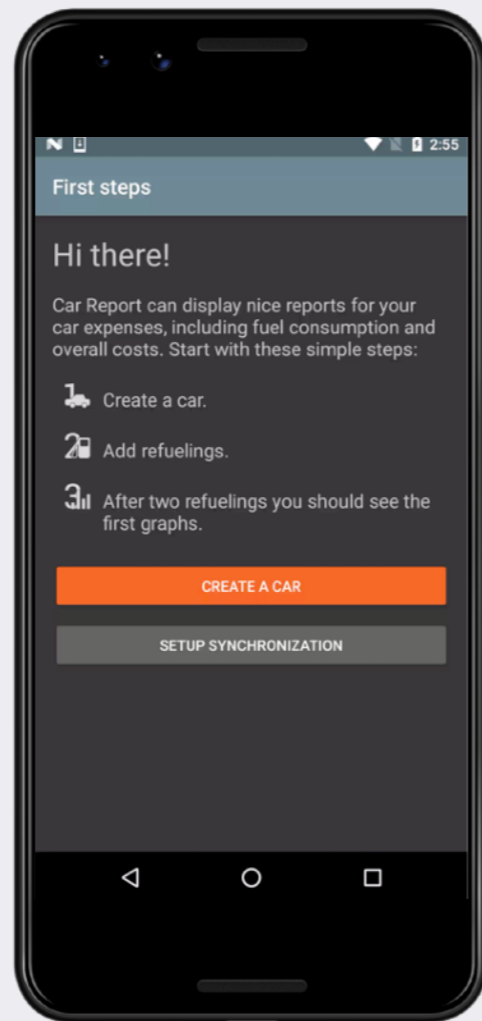
Neural Object Detection & Classification



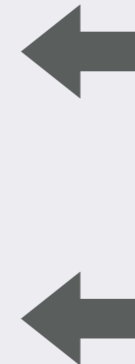
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[ 17507.045797] /dev/input/event1: 0003 0000 00000000
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[ 17507.045797] /dev/input/event1: 0003 0000 00000000
[ 17507.045797] /dev/input/event1: 0000 0000 00000000
[ 17507.078433] /dev/input/event1: 0003 0000 00000000
[ 17507.078433] /dev/input/event1: 0000 0000 00000000
[ 17509.906987] /dev/input/event1: 0003 0039 00000242
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Replay Script

OUR SOLUTION: VIDEO2SCENARIO (V2S)



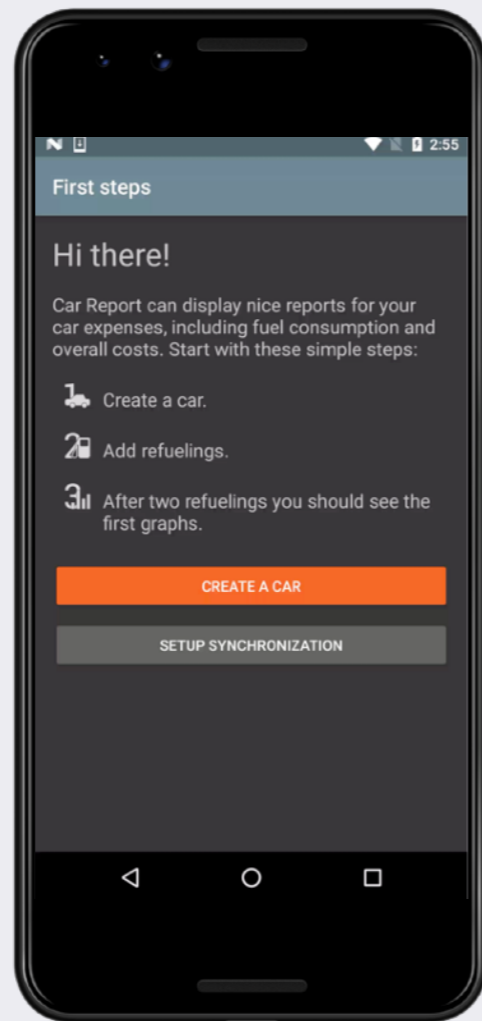
Accurate
Scenario Replay



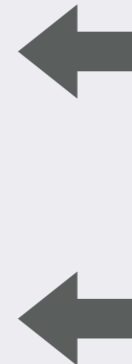
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[ 17509.906987] /dev/input/event1: 0003 0035 00000280
[ 17509.906987] /dev/input/event1: 0003 0036 0000028d
[ 17509.906987] /dev/input/event1: 0003 003a 00000030
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[ 17512.300038] /dev/input/event1: 0003 0035 000003e9
[ 17512.300038] /dev/input/event1: 0003 0036 000000b4
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[ 17512.341768] /dev/input/event1: 0003 0036 000000b1
[ 17512.341768] /dev/input/event1: 0000 0000 00000000
[ 17512.350276] /dev/input/event1: 0003 0036 000000b0
[ 17512.350276] /dev/input/event1: 0003 003a 0000002f
[ 17512.350276] /dev/input/event1: 0003 0030 00000004
[ 17512.350276] /dev/input/event1: 0000 0000 00000000
[ 17512.357952] /dev/input/event1: 0003 0036 0000000f
```

Replay Script

OUR SOLUTION: VIDEO2SCENARIO (V2S)



Accurate
Scenario Replay



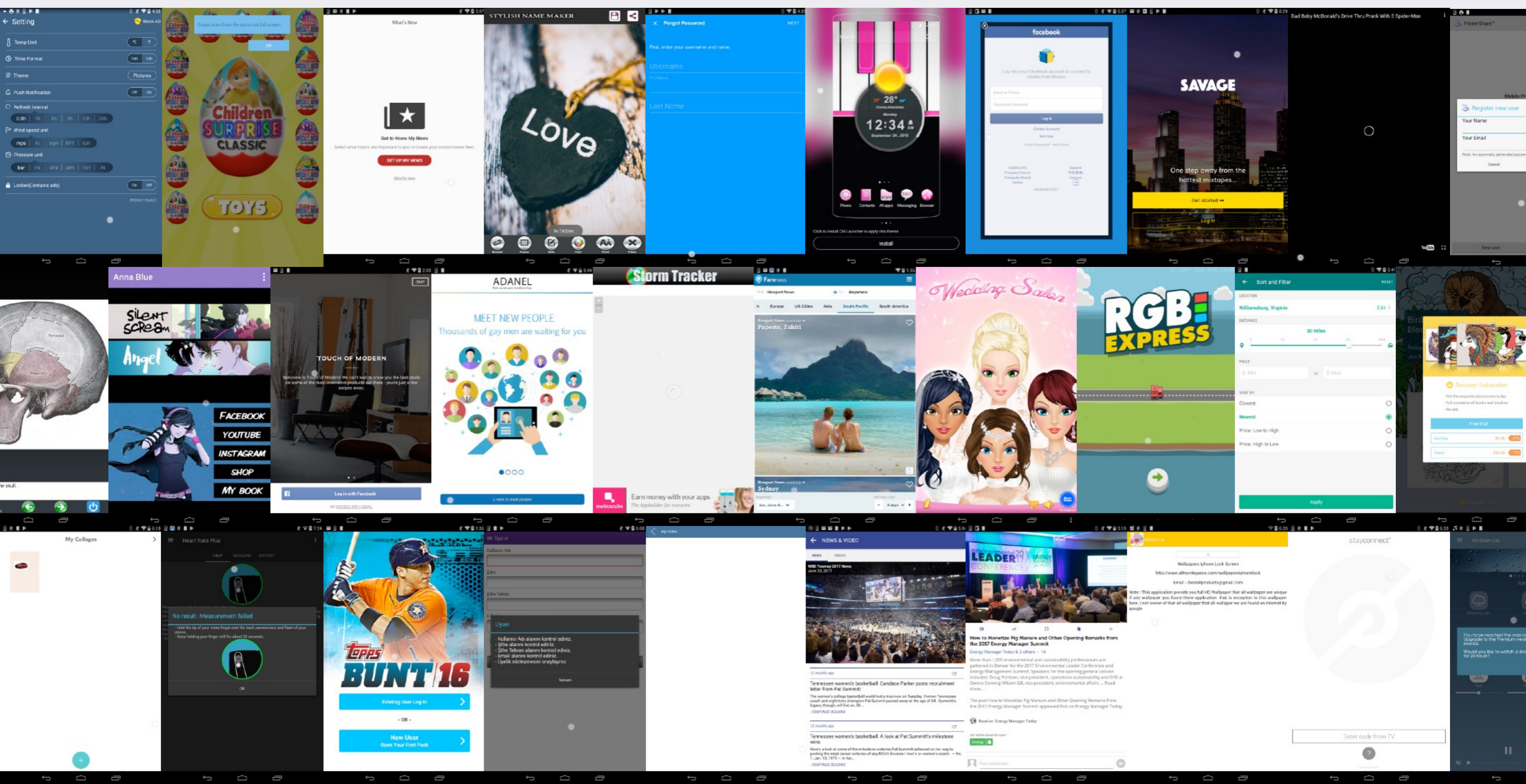
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[ 17524.482699] /dev/input/event1: 0003 0036 00000351
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[ 17524.491045] /dev/input/event1: 0003 0035 0000039e
[ 17524.491045] /dev/input/event1: 0003 0036 0000036a
[ 17524.491045] /dev/input/event1: 0000 0000 00000000
[ 17524.499417] /dev/input/event1: 0003 0035 000003a4
[ 17524.499417] /dev/input/event1: 0003 0036 00000351
[ 17524.499417] /dev/input/event1: 0000 0000 00000000
[ 17524.507811] /dev/input/event1: 0003 0035 000003a9
[ 17524.507811] /dev/input/event1: 0003 0036 0000033b
[ 17524.507811] /dev/input/event1: 0000 0000 00000000
[ 17524.516169] /dev/input/event1: 0003 0035 000003ae
[ 17524.516169] /dev/input/event1: 0003 0036 00000325
[ 17524.516169] /dev/input/event1: 0000 0000 00000000
[ 17524.524608] /dev/input/event1: 0003 0035 000003b2
[ 17524.524608] /dev/input/event1: 0003 0036 00000311
[ 17524.524608] /dev/input/event1: 0000 0000 00000000
[ 17524.532939] /dev/input/event1: 0003 0035 000003b6
[ 17524.532939] /dev/input/event1: 0003 0036 000002ff
[ 17524.532939] /dev/input/event1: 0000 0000 00000000
[ 17524.541342] /dev/input/event1: 0003 0035 000003ba
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[ 17524.549665] /dev/input/event1: 0003 0035 000003bd
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```

Replay Script

PHASE I: TOUCH DETECTION

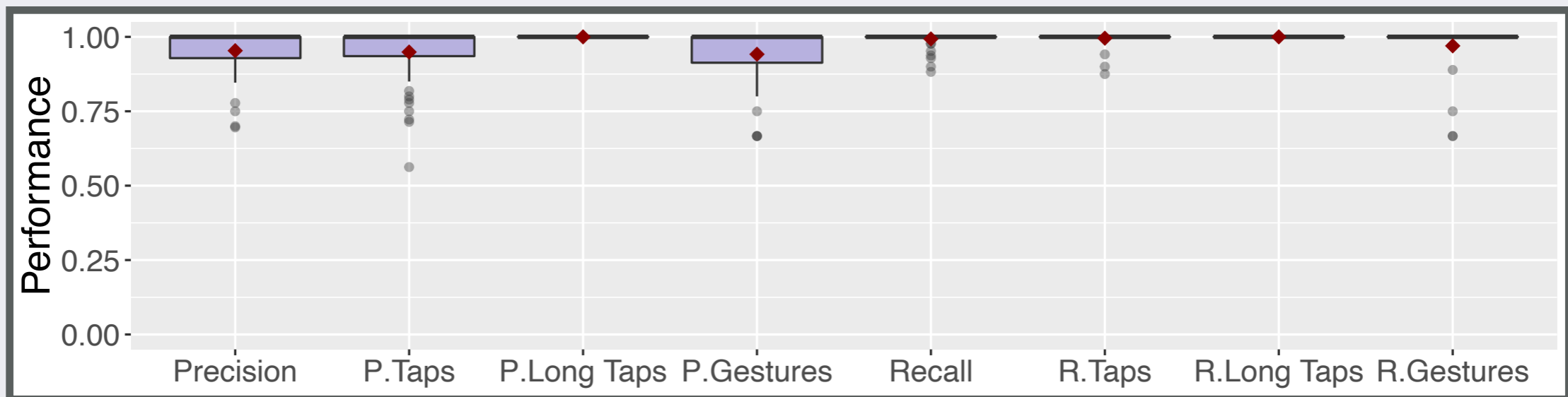


PHASE I: TOUCH DETECTION

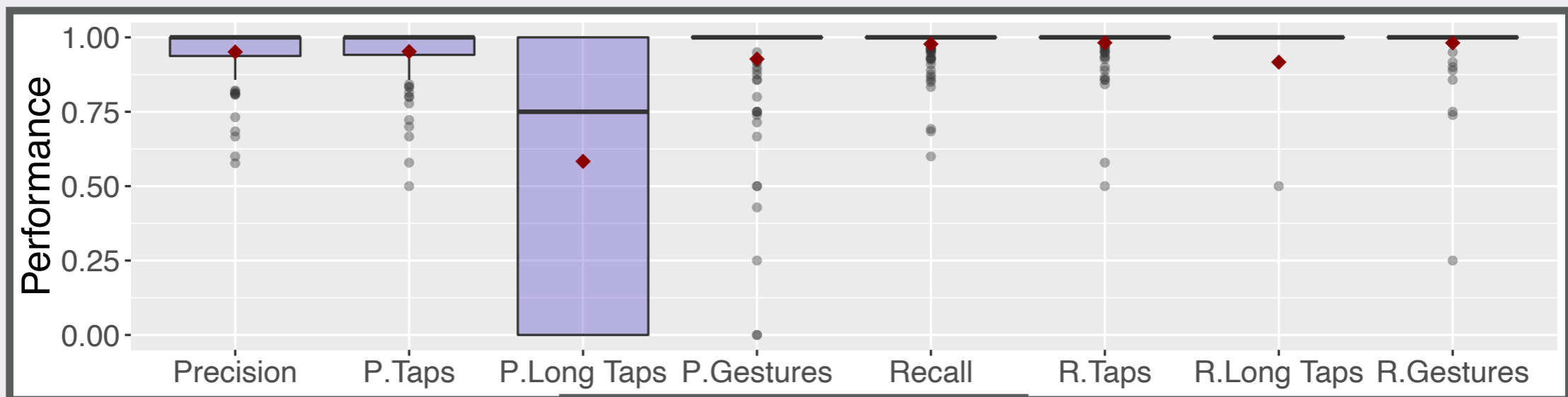


ACCURACY OF REPLAY

Scenario Replay Precision & Recall



Controlled Study



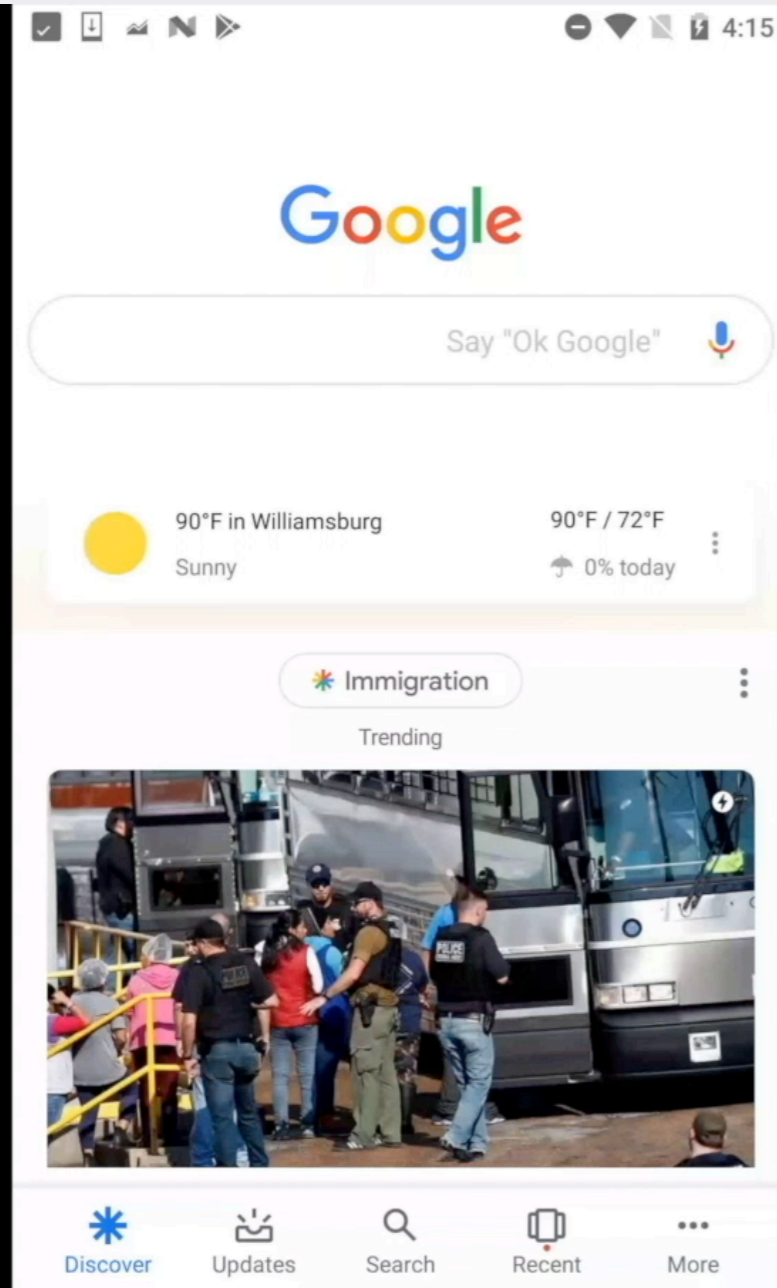
Popular Apps

ACCURACY OF REPLAY

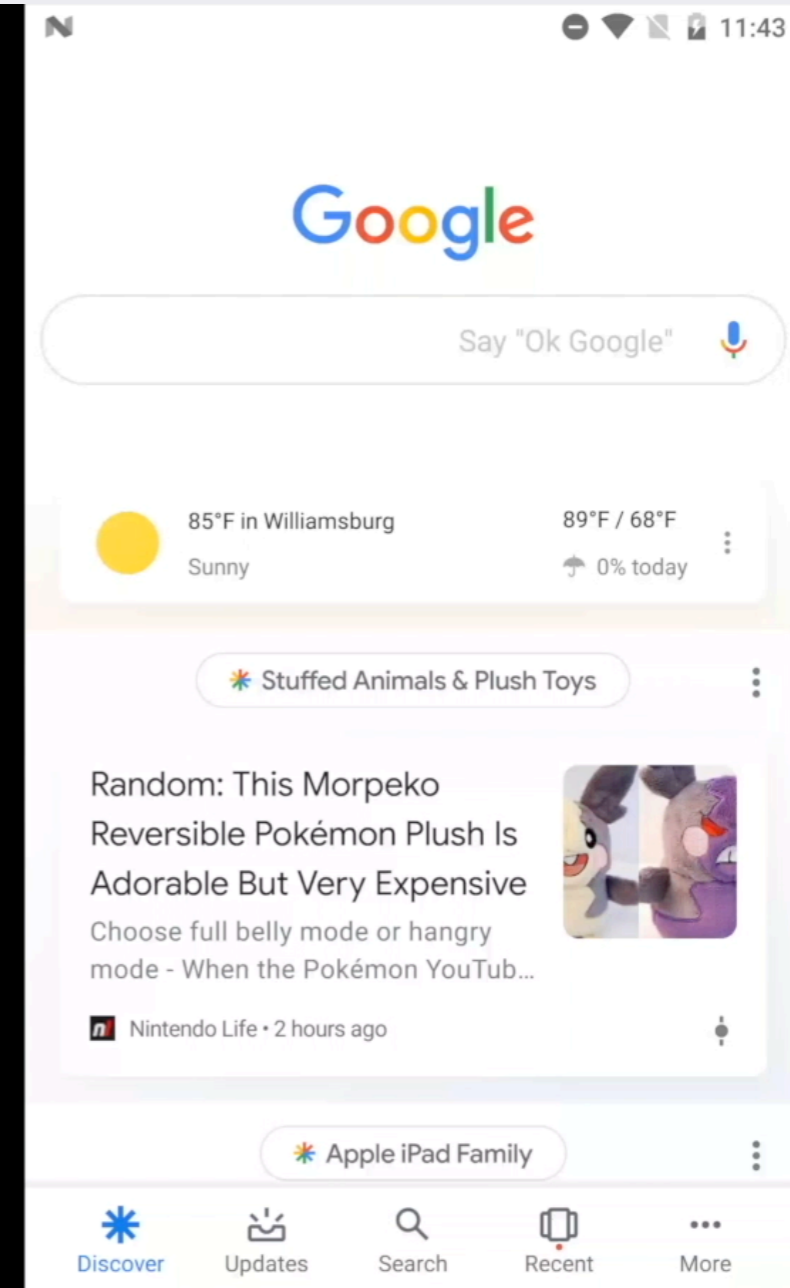
Scenario Replay Precision & Recall

Overall, precision and recall are ~95% and ~98% respectively.

RQ₃: ACCURACY OF REPLAY

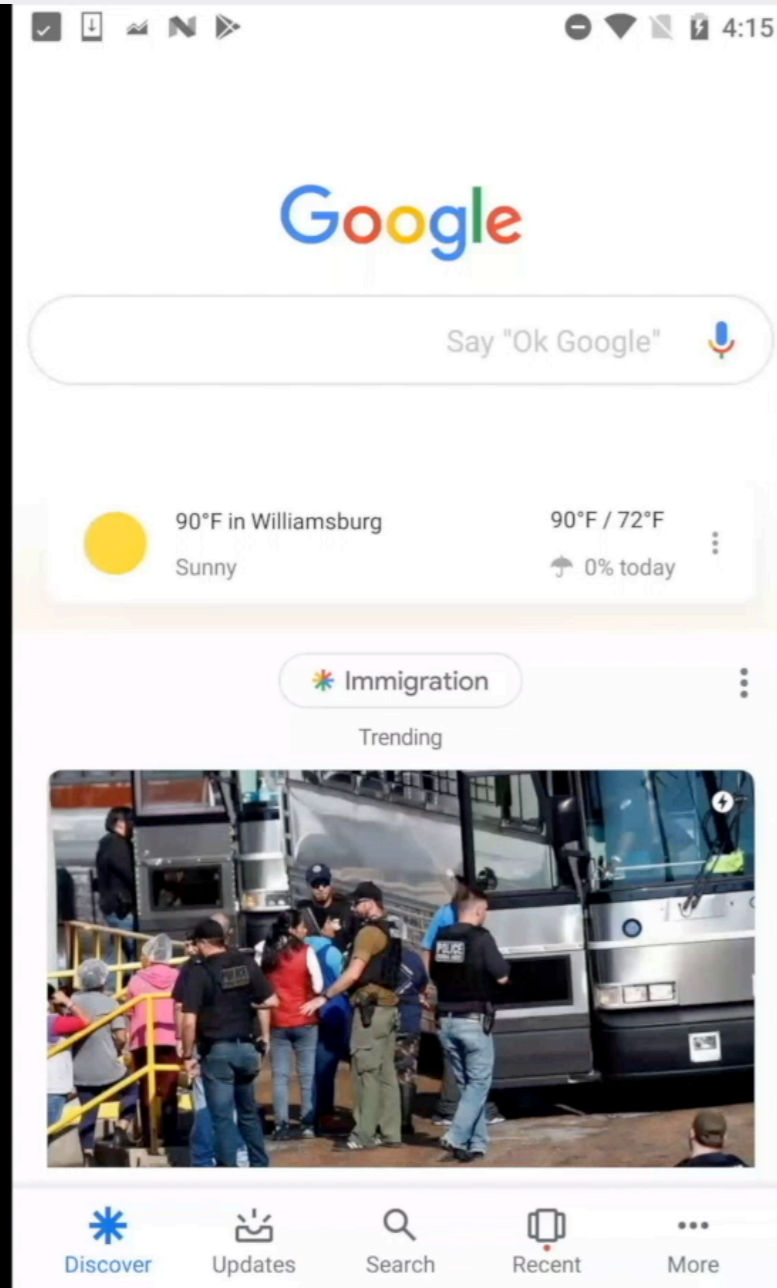


ORIGINAL

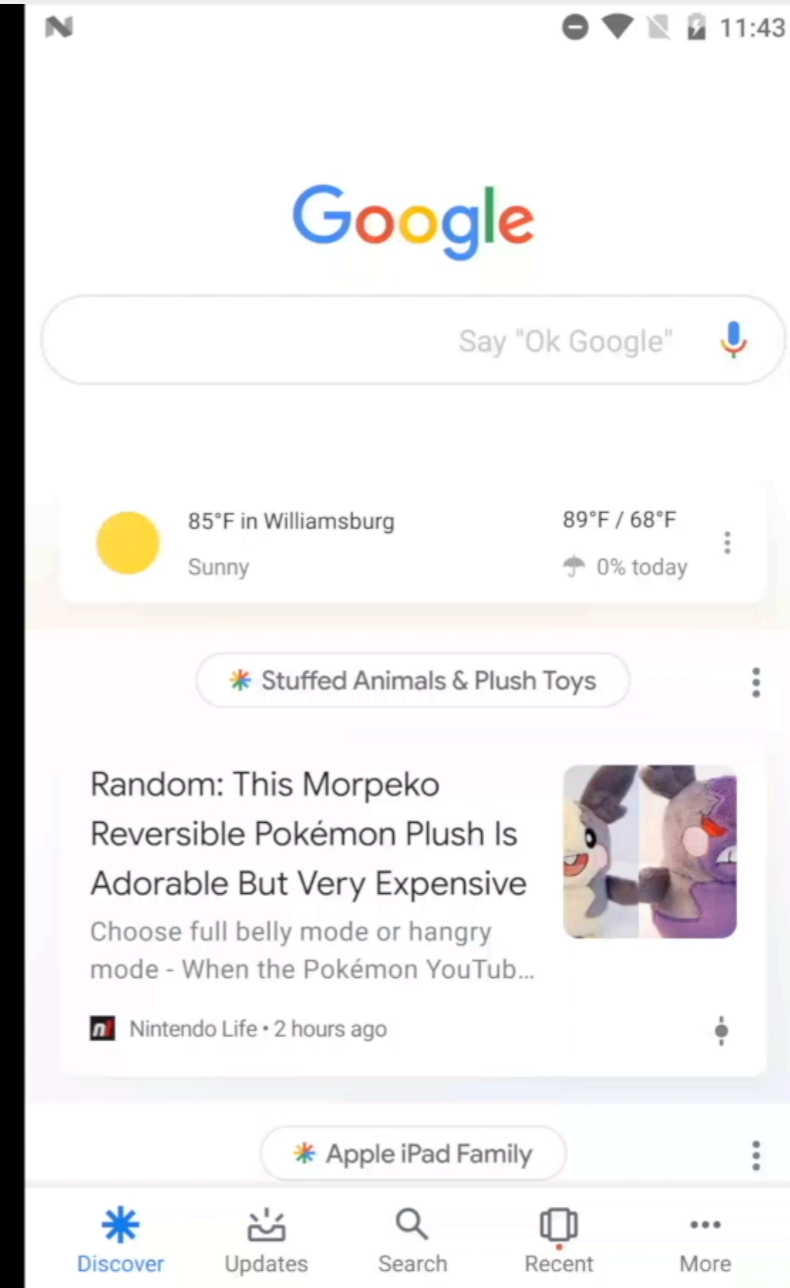


GENERATED

RQ₃: ACCURACY OF REPLAY



ORIGINAL



GENERATED

OVERVIEW OF OTHER RESEARCH AREAS & FUTURE WORK

RESEARCH AREAS & FUTURE WORK

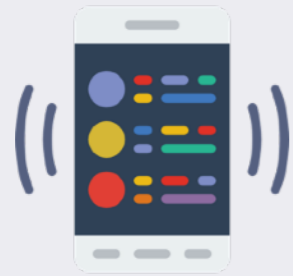
RESEARCH AREAS & FUTURE WORK

Mobile Software
Development

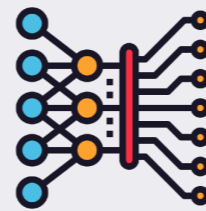


RESEARCH AREAS & FUTURE WORK

Mobile Software
Development



Deep Learning
& Software
Engineering

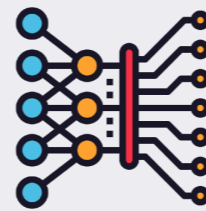


RESEARCH AREAS & FUTURE WORK

Mobile Software
Development



Deep Learning
& Software
Engineering



Transforming
Bug Reporting

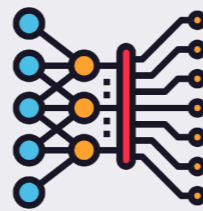


RESEARCH AREAS & FUTURE WORK

Mobile Software
Development



Deep Learning
& Software
Engineering



Transforming
Bug Reporting

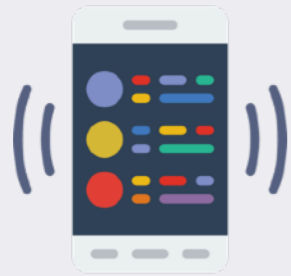


Improving Software
Traceability

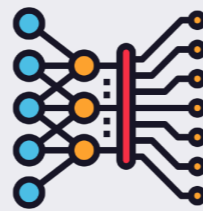


RESEARCH AREAS & FUTURE WORK

Mobile Software
Development



Deep Learning
& Software
Engineering



Transforming
Bug Reporting



Improving Software
Traceability



Developer-Centric
Software Security

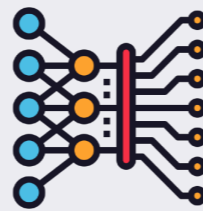


RESEARCH AREAS & FUTURE WORK

Mobile Software
Development



Deep Learning
& Software
Engineering



Transforming
Bug Reporting



Improving Software
Traceability



Developer-Centric
Software Security



Automating
Development of
Accessibility Features



RESEARCH AREAS & FUTURE WORK

Applied Techniques:

Machine Learning
Natural Language Processing
Computer Vision
Program Analysis

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SWE Research Group



Dr. Kevin Moran
Assistant Professor

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for Machine Learning

Academic Collaborators



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Industrial Collaborators



facebook



Thank you!



Kevin Moran
Assistant Professor
kpmoran@gmu.edu
<https://www.kpmoran.com>



Feel free to come chat (virtually)!

ADDITIONAL SLIDES
