

# GitHub and Deep Learning on Graphs of Code

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# Outline



- Introduction to problem: detection of duplicate code
- Primer on types of duplicate code
- Our approach: machine learning on Abstract Syntax Trees (ASTs)
- Results
- Future work

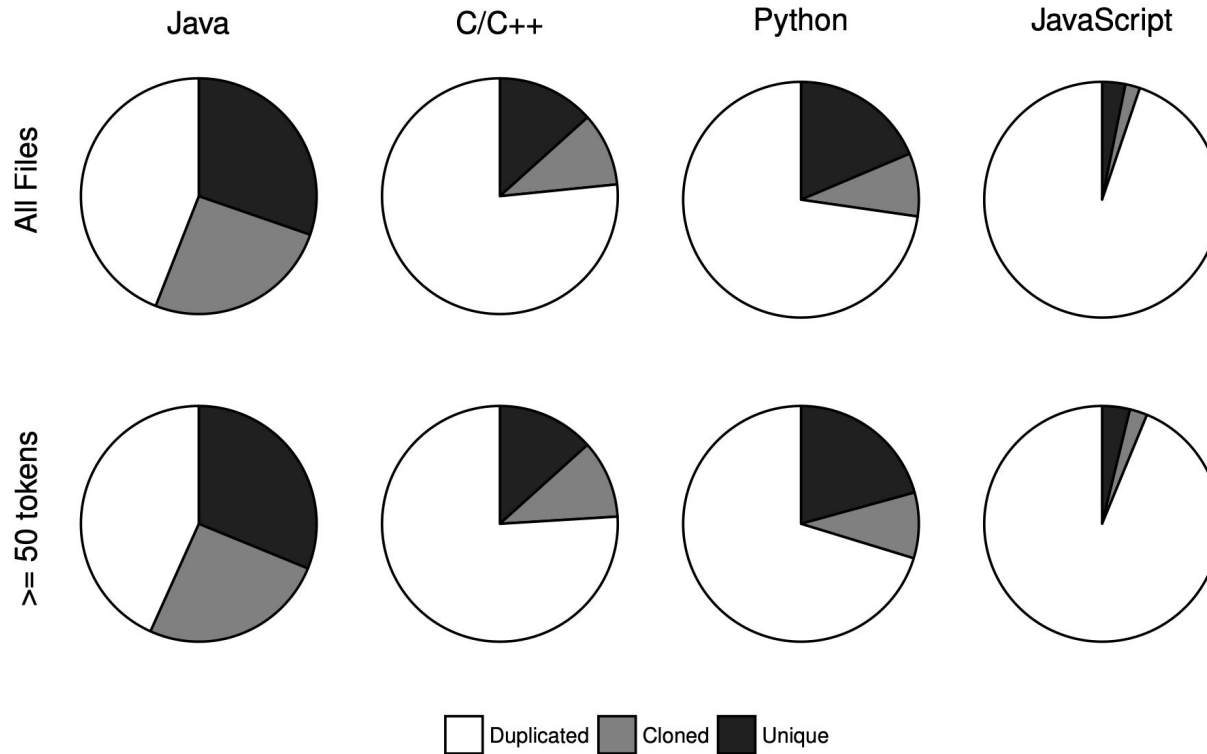


Fig. 5. File-level duplication for entire dataset and excluding small files.

[DejuVu: A Map of Code Duplication on GitHub](#) (Lopes et al., *Proc. ACM Program. Lang.* (1), 2017)

# Code Duplication Primer



- **Type 0**: completely identical
- **Type 1**: only difference is in comments and whitespace
- **Type 2**: can also include variations in identifier names and literal values
- **Type 3**: syntactically similar with differences as the statement level (can be added, removed, or modified)
- **Type 4**: syntactically different but semantically similar

# Code Duplication Primer



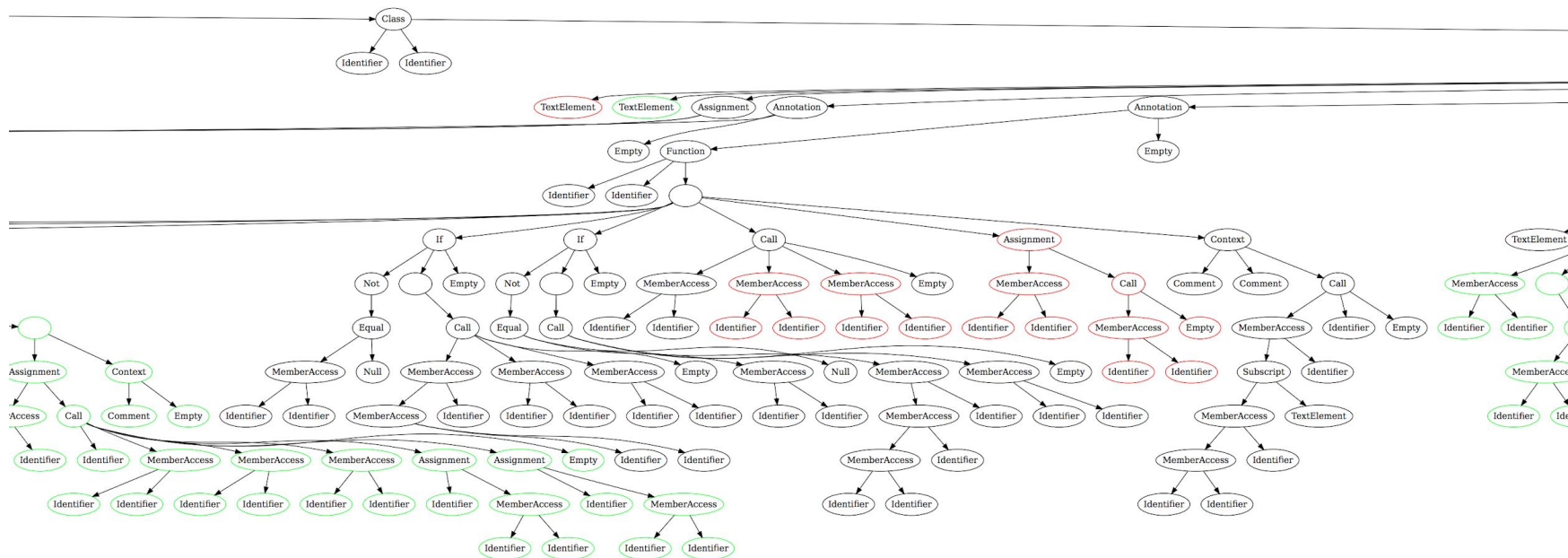
- **Type 0**: completely identical
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# How similar are these functions?

```
def foo(x, y):  
    return x + y
```

```
def bar(z):  
    z = z+1  
    return z
```





# ASTs to Graphs



```
[{'graph': {'edges': [{'source': 0, 'target': 1},
{'source': 1, 'target': 2},
{'source': 1, 'target': 10},
{'source': 2, 'target': 3},
...
{'source': 7, 'target': 8},
{'source': 7, 'target': 9}],
'vertices': [{'id': 0,
'sourceRange': [1, 33],
'sourceSpan': {'end': [4, 1], 'start': [2, 1]},
'term': 'Statements'},
{'id': 1,
'sourceRange': [1, 32],
'sourceSpan': {'end': [3, 17], 'start': [2, 1]},
'term': 'Annotation'},
{'id': 2,
'sourceRange': [1, 32],
'sourceSpan': {'end': [3, 17], 'start': [2, 1]},
'term': 'Function'},
...
{'id': 8,
'sourceRange': [27, 28],
'sourceSpan': {'end': [3, 13], 'start': [3, 12]},
'term': 'Identifier'},
{'id': 9,
'sourceRange': [31, 32],
'sourceSpan': {'end': [3, 17], 'start': [3, 16]},
'term': 'Identifier'},
{'id': 10,
'sourceRange': [1, 32],
'sourceSpan': {'end': [3, 17], 'start': [2, 1]},
'term': 'Empty'}}],
'language': 'Python',
'path': 'example1.py'}}
```



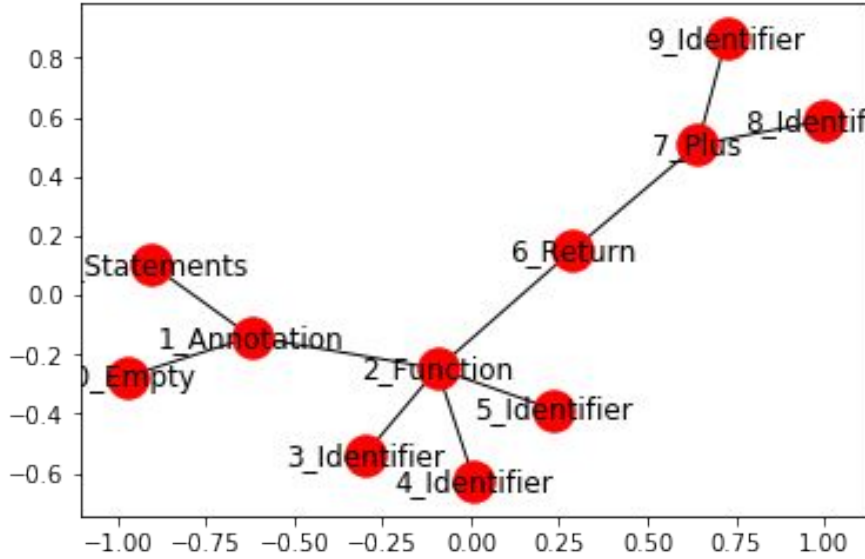
```
def foo(x, y):
    return x + y
```

```
def bar(z):
    z = z+1
    return z
```



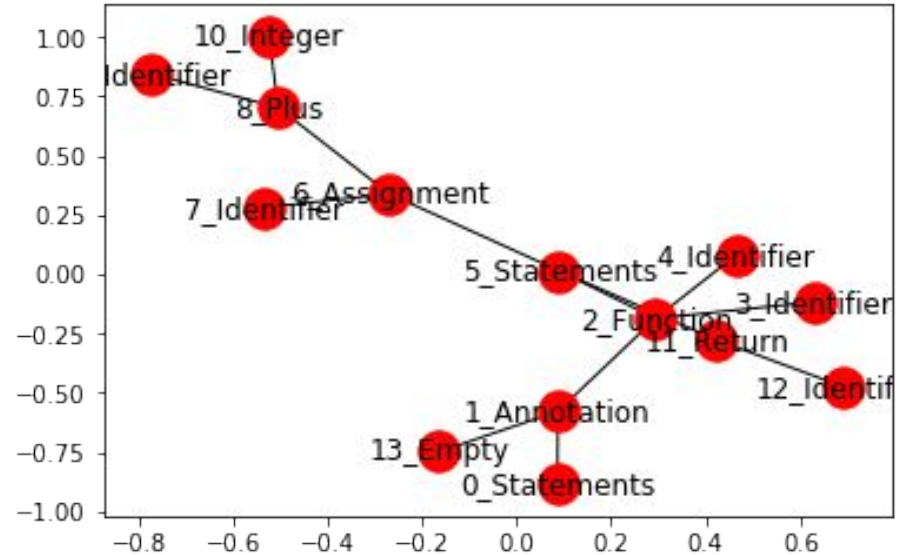
```
[{'graph': {'edges': [{'source': 0, 'target': 1},
{'source': 1, 'target': 2},
{'source': 1, 'target': 13},
{'source': 2, 'target': 3},
...
{'source': 6, 'target': 8},
{'source': 8, 'target': 9},
{'source': 8, 'target': 10},
{'source': 11, 'target': 12}],
'vertices': [{'id': 0,
'sourceRange': [1, 38],
'sourceSpan': {'end': [5, 1], 'start': [2, 1]},
'term': 'Statements'},
{'id': 1,
'sourceRange': [1, 37],
'sourceSpan': {'end': [4, 13], 'start': [2, 1]},
'term': 'Annotation'},
{'id': 2,
'sourceRange': [1, 37],
'sourceSpan': {'end': [4, 13], 'start': [2, 1]},
'term': 'Function'},
...
{'id': 11,
'sourceRange': [29, 37],
'sourceSpan': {'end': [4, 13], 'start': [4, 5]},
'term': 'Return'},
{'id': 12,
'sourceRange': [36, 37],
'sourceSpan': {'end': [4, 13], 'start': [4, 12]},
'term': 'Identifier'},
{'id': 13,
'sourceRange': [1, 37],
'sourceSpan': {'end': [4, 13], 'start': [2, 1]},
'term': 'Empty'}}],
'language': 'Python',
'path': 'example2.py'}}
```

# ASTs to Graphs (cont.)



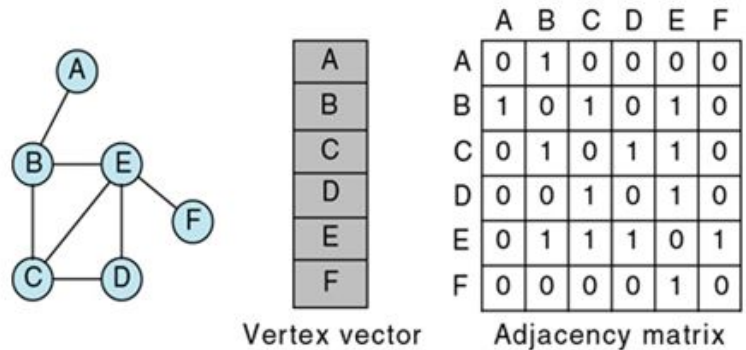
```
def foo(x, y):  
    return x + y
```

```
def bar(z):  
    z = z+1  
    return z
```

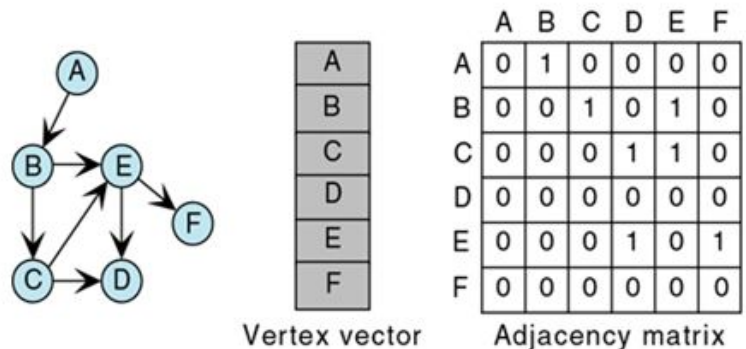




Graph calculations are typically done through the adjacency matrix...



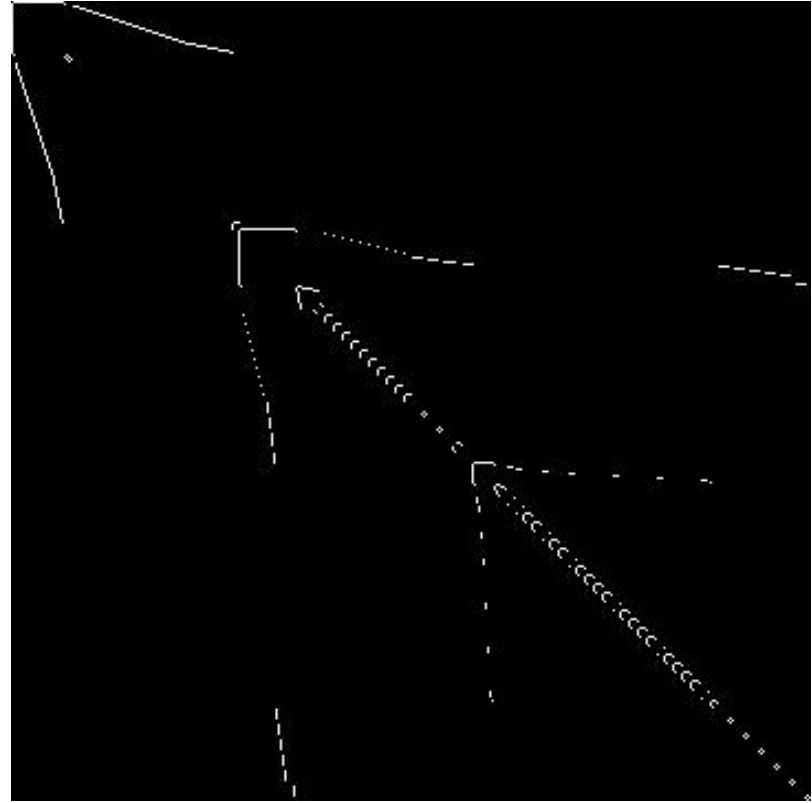
(a) Adjacency matrix for non-directed graph



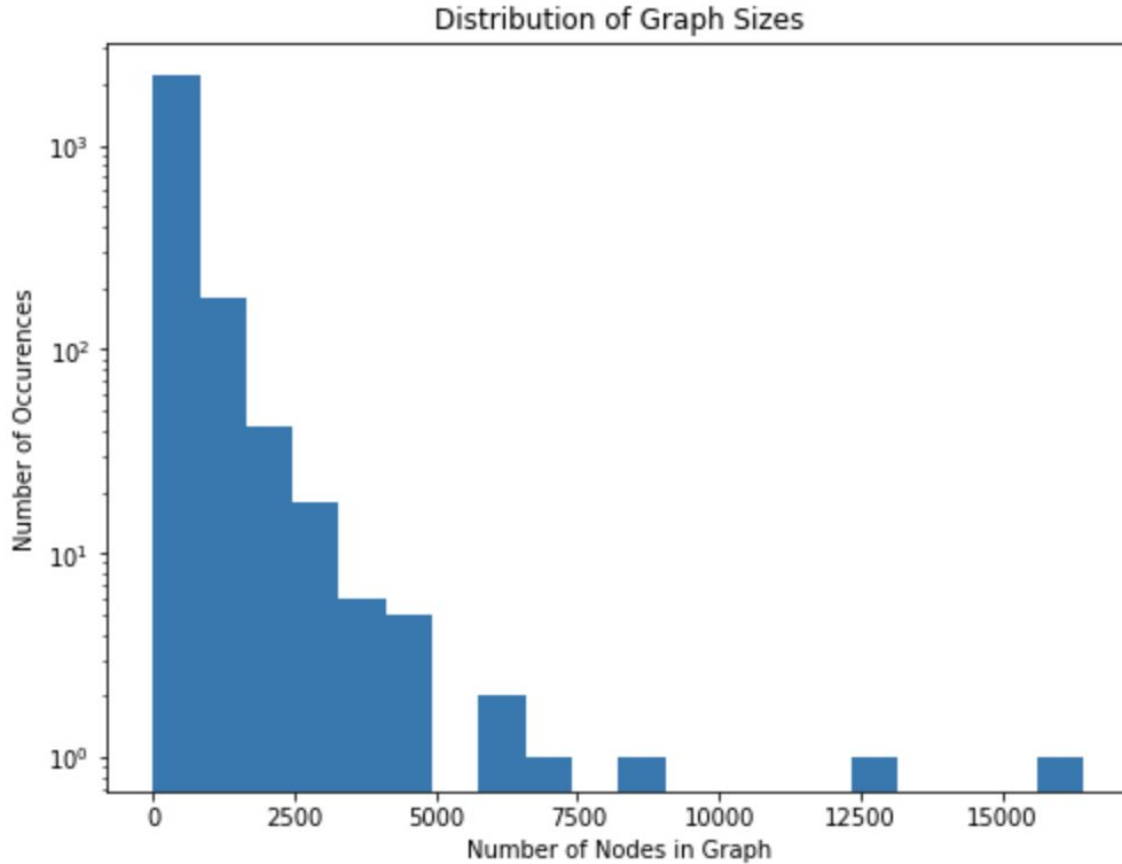
(a) Adjacency matrix for directed graph



...which, really, are just  
(binary) images!



# Adjacency Matrix Sizes



# General Workflow

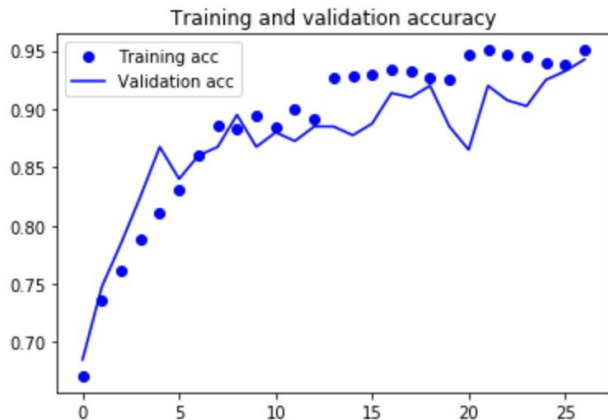


- Identify two files or functions with potential clone from metadata files
- Convert adjacency JSON to images
- Combine adjacency images two a single “clone image”
- Model!
  - file : file
  - function : function
  - Keras convolutional neural network

# Sample file:file Clone Image



# file:file Model Results



## Total numbers of files:

Total training neg: 5000

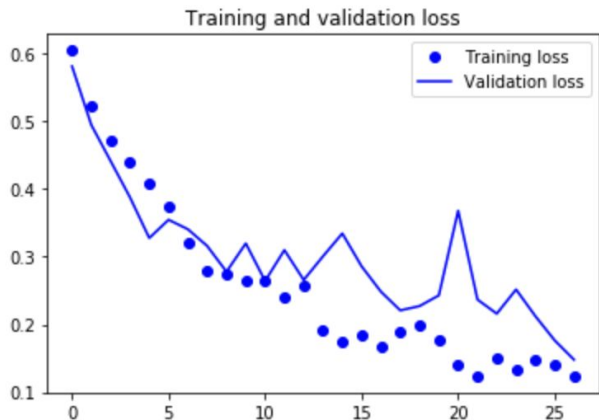
Total training type3: 8220

Total validation neg: 2500

Total validation type3: 4214

Total test neg: 2500

Total test type3: 4126

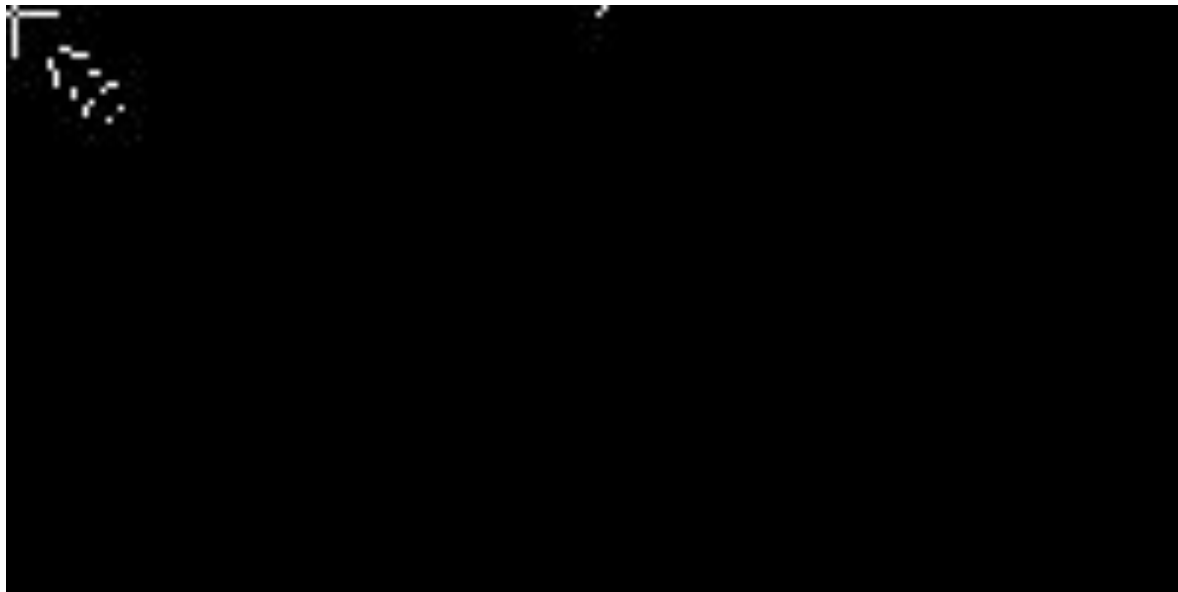


**Test accuracy: 0.925**

**Test loss: 0.210**



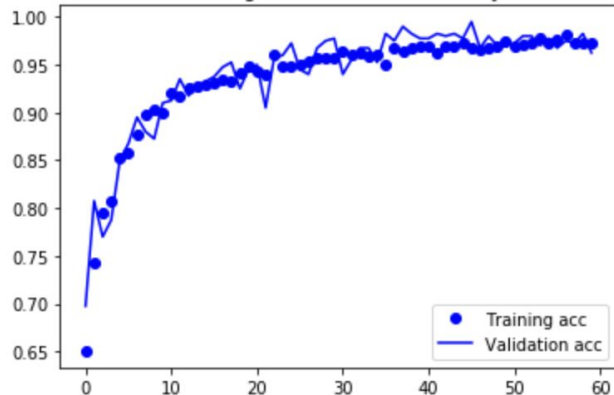
# Sample function:function Clone Image



# function:function Model Results



Training and validation accuracy



## Total numbers of files:

Total training neg: 27966

Total training type3: 24909

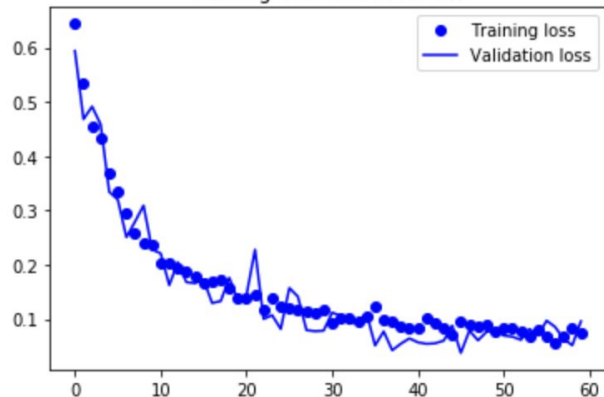
Total validation neg: 13982

Total validation type3: 12454

Total test neg: 13984

Total test type3: 12454

Training and validation loss



**Test accuracy: 0.977**

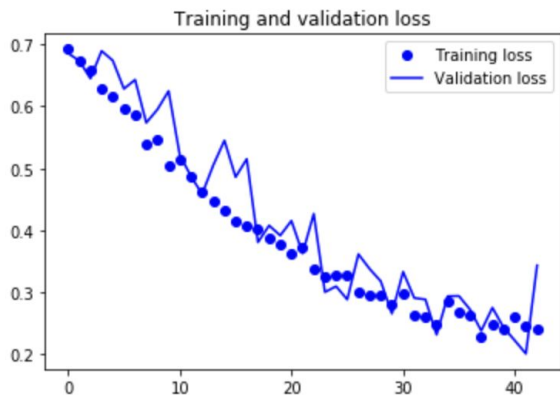
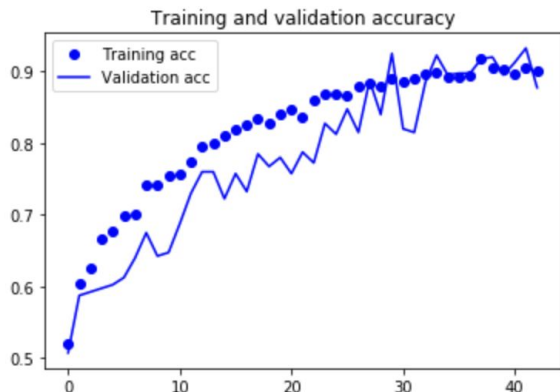
**Test loss: 0.066**



# Adding Noise (1%)



# function:function Noisy Model Results



## Total numbers of files:

Total training neg: 27966

Total training type3: 24909

Total validation neg: 13982

Total validation type3: 12454

Total test neg: 13984

Total test type3: 12454

**Test accuracy: 0.888**

**Test loss: 0.349**



- Clone detection
  - Current approach is  $O(n^2)$ 
    - How to get out of this?
  - Expand out to Type 4
    - With what training set?
- Graphs in general
  - All of GitHub data can be represented as a graph!

# GRAPH ALL THE THINGS



# Thank You!



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