

Exploring Nondeterminism in Tools That Automatically Generate or Analyse Test Suites

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Installing and loading packages

```
install.packages("ggpubr")
```

```
## Installing package into '/home/beatriz/R/x86_64-pc-linux-gnu-library/3.4'  
## (as 'lib' is unspecified)
```

```
install.packages("ggplot2")
```

```
## Installing package into '/home/beatriz/R/x86_64-pc-linux-gnu-library/3.4'  
## (as 'lib' is unspecified)
```

```
install.packages("reshape")
```

```
## Installing package into '/home/beatriz/R/x86_64-pc-linux-gnu-library/3.4'  
## (as 'lib' is unspecified)
```

```
install.packages("reshape2")
```

```
## Installing package into '/home/beatriz/R/x86_64-pc-linux-gnu-library/3.4'  
## (as 'lib' is unspecified)
```

```
install.packages("dplyr")
```

```
## Installing package into '/home/beatriz/R/x86_64-pc-linux-gnu-library/3.4'  
## (as 'lib' is unspecified)
```

```
library("ggplot2")  
library("reshape")  
library("reshape2")
```

```
##  
## Attaching package: 'reshape2'
```

```
## The following objects are masked from 'package:reshape':  
##  
##   colsplit, melt, recast
```

```
library("ggpubr")
```

```
## Loading required package: magrittr
```

```
library("dplyr")
```

```
##  
## Attaching package: 'dplyr'
```

```
## The following object is masked from 'package:reshape':  
##  
##   rename
```

```
## The following objects are masked from 'package:stats':  
##  
##   filter, lag
```

```
## The following objects are masked from 'package:base':  
##  
##   intersect, setdiff, setequal, union
```

Reading data

```
evo <- read.csv('/home/beatriz/Documents/SRC-2019/statistics/exploringNondeterminism/
evo1Execution.csv', sep = ",", header = TRUE)
evoMelted <- melt(evo, id = c("Project", "Tool"))
#evoMelted

randoop <- read.csv('/home/beatriz/Documents/SRC-2019/statistics/exploringNondeterminism/
randoop1Execution.csv', sep = ",", header = TRUE)
randoopMelted <- melt(randoop, id = c("Project", "Tool"))
#randoopMelted

pitEvo <- read.csv('/home/beatriz/Documents/SRC-2019/statistics/exploringNondeterminism/
pitSameProjectSameTestSuiteEvo.csv', sep = ",", header = TRUE)
pitEvoMelted <- melt(pitEvo, id = c("Project", "Tool"))
#pitEvoMelted

pitRandoop <- read.csv('/home/beatriz/Documents/SRC-2019/statistics/exploringNondeterminism/
pitSameProjectSameTestSuiteRandoop.csv', sep = ",", header = TRUE)
pitRandoopMelted <- melt(pitRandoop, id = c("Project", "Tool"))
#pitRandoopMelted

pitManual <- read.csv('/home/beatriz/Documents/SRC-2019/statistics/exploringNondeterminism/
pitSameProjectSameTestSuiteManual.csv', sep = ",", header = TRUE)
pitManualMelted <- melt(pitManual, id = c("Project", "Tool"))
#pitManualMelted
```

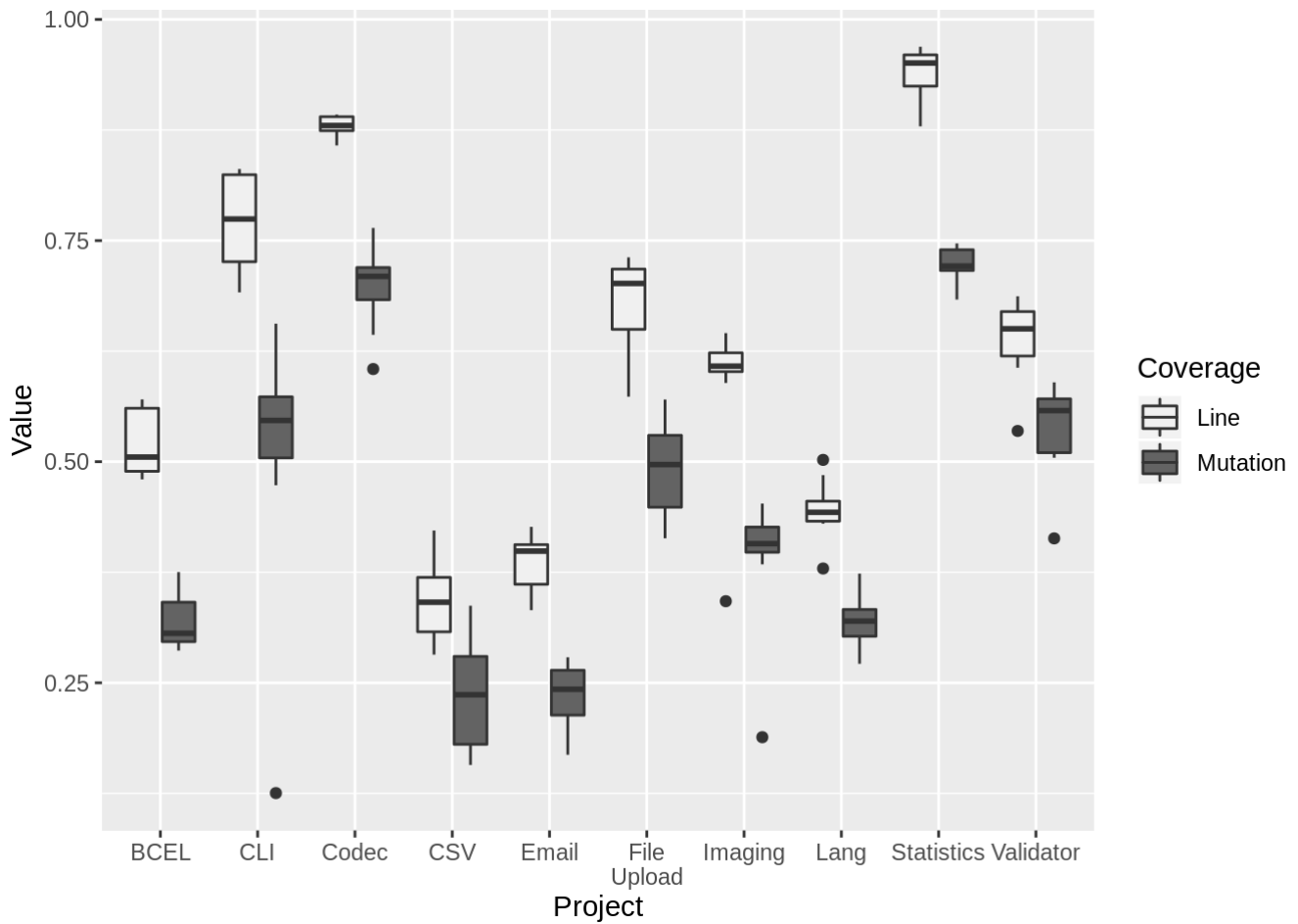
Declaring common values

```
projects=c("BCEL","CLI","Codec", "CSV", "Email", "File\nUpload", "Imaging", "Lang",
"Statistics", "Validator")
```

PIT was executed once for each test suite.

Here we analyse the difference between test suites generated for the same project with EvoSuite in terms of line and mutation coverage.

```
ggplot(data = evoMelted, aes(x=Project, y=value)) +
  geom_boxplot(aes(fill=variable)) +
  scale_x_discrete(labels=projects) +
  labs(x = "Project", y = "Value") +
  scale_fill_manual(values = c("#f0f0f0", "#636363"), name = "Coverage",
                    labels = c("Line", "Mutation"))
```



```
# Line Coverage mean, variation and standard deviation
```

```
evoMeanVariationSDLC <- evo %>% group_by(Project) %>% summarise(LineCoverageAverage =
mean(LineCoverage),
                                                                    LineCoverageVariation =
var(LineCoverage),
                                                                    LineCoverageStandardDev
iation = sd(LineCoverage))
evoMeanVariationSDLC
```

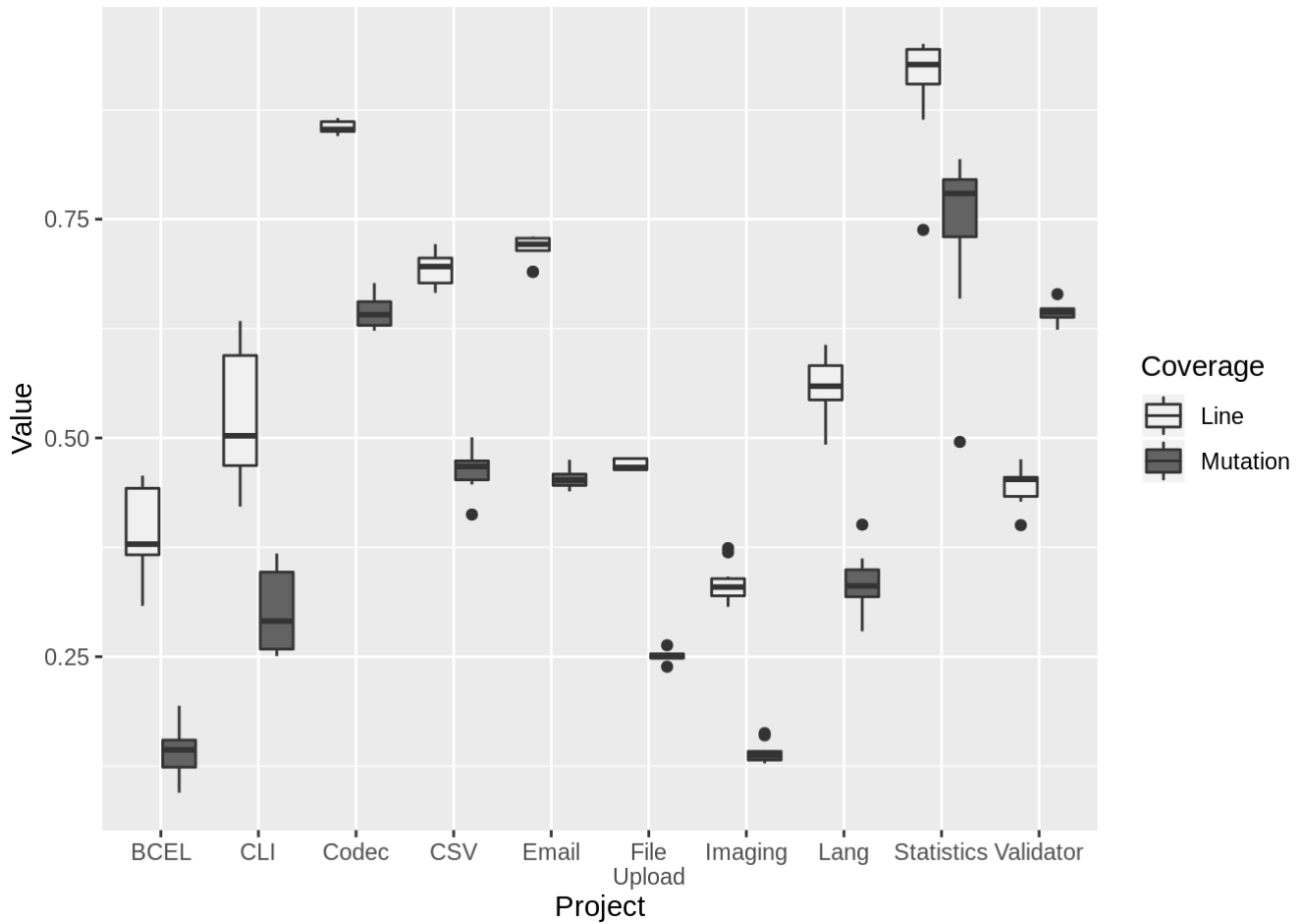
```
## # A tibble: 10 x 4
##   Project LineCoverageAvera... LineCoverageVariat... LineCoverageStandardDe...
##   <fct>          <dbl>          <dbl>          <dbl>
## 1 bcel          0.520          0.00147        0.0383
## 2 cli           0.772          0.00275        0.0524
## 3 codec         0.879          0.000141       0.0119
## 4 csv           0.344          0.00221        0.0470
## 5 email         0.384          0.00115        0.0339
## 6 fileuplo...  0.681          0.00258        0.0508
## 7 imaging       0.587          0.00769        0.0877
## 8 lang          0.446          0.00109        0.0331
## 9 statisti...  0.938          0.00111        0.0334
## 10 validator    0.639          0.00210        0.0458
```

```
# Mutation Coverage mean, variation and standard deviation
evoMeanVariationSDMC <- evo %>% group_by(Project) %>% summarise(MutationCoverageAvera
ge = mean(MutationCoverage),
                                                                MutationCoverageVariati
on = var(MutationCoverage),
                                                                MutationCoverageStandar
dDeviation = sd(MutationCoverage))
evoMeanVariationSDMC
```

```
## # A tibble: 10 x 4
##   Project    MutationCoverageAv... MutationCoverageVa... MutationCoverageStand...
##   <fct>          <dbl>          <dbl>          <dbl>
## 1 bcel          0.319          0.000983       0.0314
## 2 cli           0.510          0.0207        0.144
## 3 codec         0.696          0.00205       0.0453
## 4 csv           0.239          0.00423       0.0650
## 5 email         0.234          0.00148       0.0385
## 6 fileuplo...   0.491          0.00299       0.0547
## 7 imaging       0.391          0.00547       0.0740
## 8 lang          0.318          0.000806      0.0284
## 9 statisti...   0.724          0.000356      0.0189
## 10 validator    0.536          0.00276       0.0525
```

Here we analyse the difference between test suites generated for the same project with Randoop in terms of line and mutation coverage.

```
ggplot(data = randoopMelted, aes(x=Project, y=value)) +
  geom_boxplot(aes(fill=variable)) +
  scale_x_discrete(labels=projects) +
  labs(x ="Project", y = "Value") +
  scale_fill_manual(values = c("#f0f0f0", "#636363"), name = "Coverage",
                    labels = c("Line", "Mutation"))
```



```
# Line Coverage mean, variation and standard deviation
randoopMeanVariationSDLC <- randoop %>% group_by(Project) %>% summarise(LineCoverageA
verage = mean(LineCoverage),
                                                                    LineCoverageVariation =
var(LineCoverage),
                                                                    LineCoverageStandardDev
iation = sd(LineCoverage))
randoopMeanVariationSDLC
```

```
## # A tibble: 10 x 4
##   Project LineCoverageAvera... LineCoverageVariat... LineCoverageStandardDe...
##   <fct>          <dbl>          <dbl>          <dbl>
## 1 bcel          0.389          0.00294        0.0543
## 2 cli           0.524          0.00537        0.0733
## 3 codec         0.855          0.0000472      0.00687
## 4 csv           0.692          0.000353       0.0188
## 5 email         0.717          0.000233       0.0153
## 6 fileuplo...   0.469          0.0000431      0.00657
## 7 imaging        0.334          0.000488       0.0221
## 8 lang          0.559          0.00109        0.0330
## 9 statisti...   0.905          0.00417        0.0646
## 10 validator    0.447          0.000507       0.0225
```

```
# Mutation Coverage mean, variation and standard deviation
randoopMeanVariationSDMC <- randoop %>% group_by(Project) %>% summarise(MutationCover
ageAverage = mean(MutationCoverage),
                                MutationCoverageVariati
on = var(MutationCoverage),
                                MutationCoverageStandar
dDeviation = sd(MutationCoverage))
randoopMeanVariationSDMC
```

```
## # A tibble: 10 x 4
##   Project MutationCoverageAv... MutationCoverageVa... MutationCoverageStand...
##   <fct>          <dbl>          <dbl>          <dbl>
## 1 bcel          0.142          0.000864       0.0294
## 2 cli           0.303          0.00226       0.0475
## 3 codec         0.643          0.000324       0.0180
## 4 csv           0.465          0.000650       0.0255
## 5 email         0.454          0.000123       0.0111
## 6 fileuplo...   0.251          0.0000393      0.00627
## 7 imaging       0.140          0.000143       0.0119
## 8 lang          0.334          0.00108       0.0329
## 9 statisti...   0.741          0.00977       0.0989
## 10 validator    0.643          0.000116       0.0108
```

Here we compare the standard deviations of line and mutation coverage of the test suites generated with EvoSuite and Randoop.

```
# Line Coverage
wilcox.test(evoMeanVariationSDLC$LineCoverageStandardDeviation, randoopMeanVariationS
DLC$LineCoverageStandardDeviation, paired = TRUE)
```

```
##
## Wilcoxon signed rank test
##
## data:  evoMeanVariationSDLC$LineCoverageStandardDeviation and randoopMeanVariation
SDLC$LineCoverageStandardDeviation
## V = 39, p-value = 0.2754
## alternative hypothesis: true location shift is not equal to 0
```

```
# Mutation Coverage
wilcox.test(evoMeanVariationSDMC$MutationCoverageStandardDeviation, randoopMeanVariat
ionSDMC$MutationCoverageStandardDeviation, paired = TRUE)
```

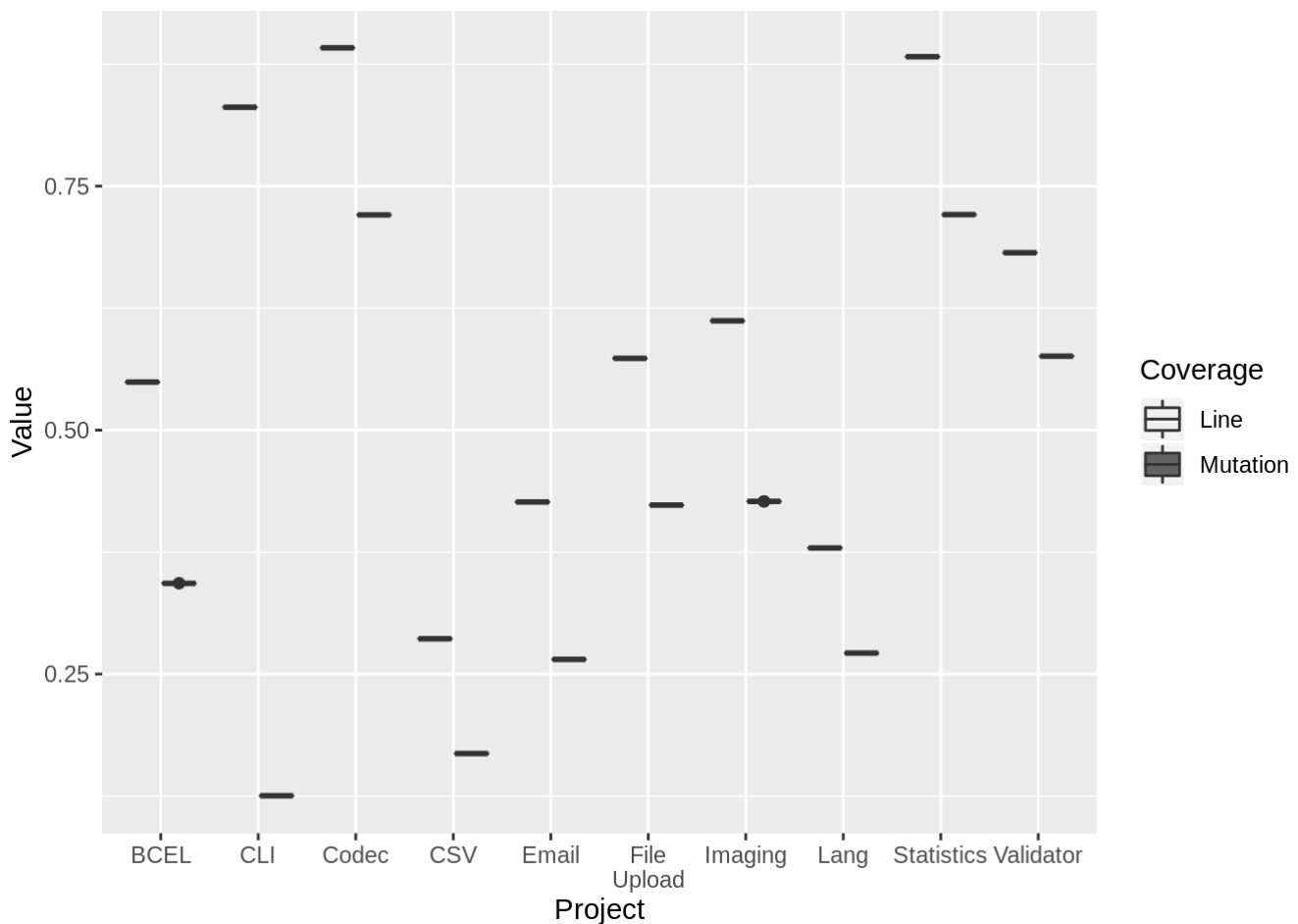
```
##
## Wilcoxon signed rank test
##
## data:  evoMeanVariationSDMC$MutationCoverageStandardDeviation and randoopMeanVaria
tionSDMC$MutationCoverageStandardDeviation
## V = 44, p-value = 0.1055
## alternative hypothesis: true location shift is not equal to 0
```

PIT was executed 10 times for each test suite.

Here we analyse PIT's outputs when executed for the same project and the same test suites

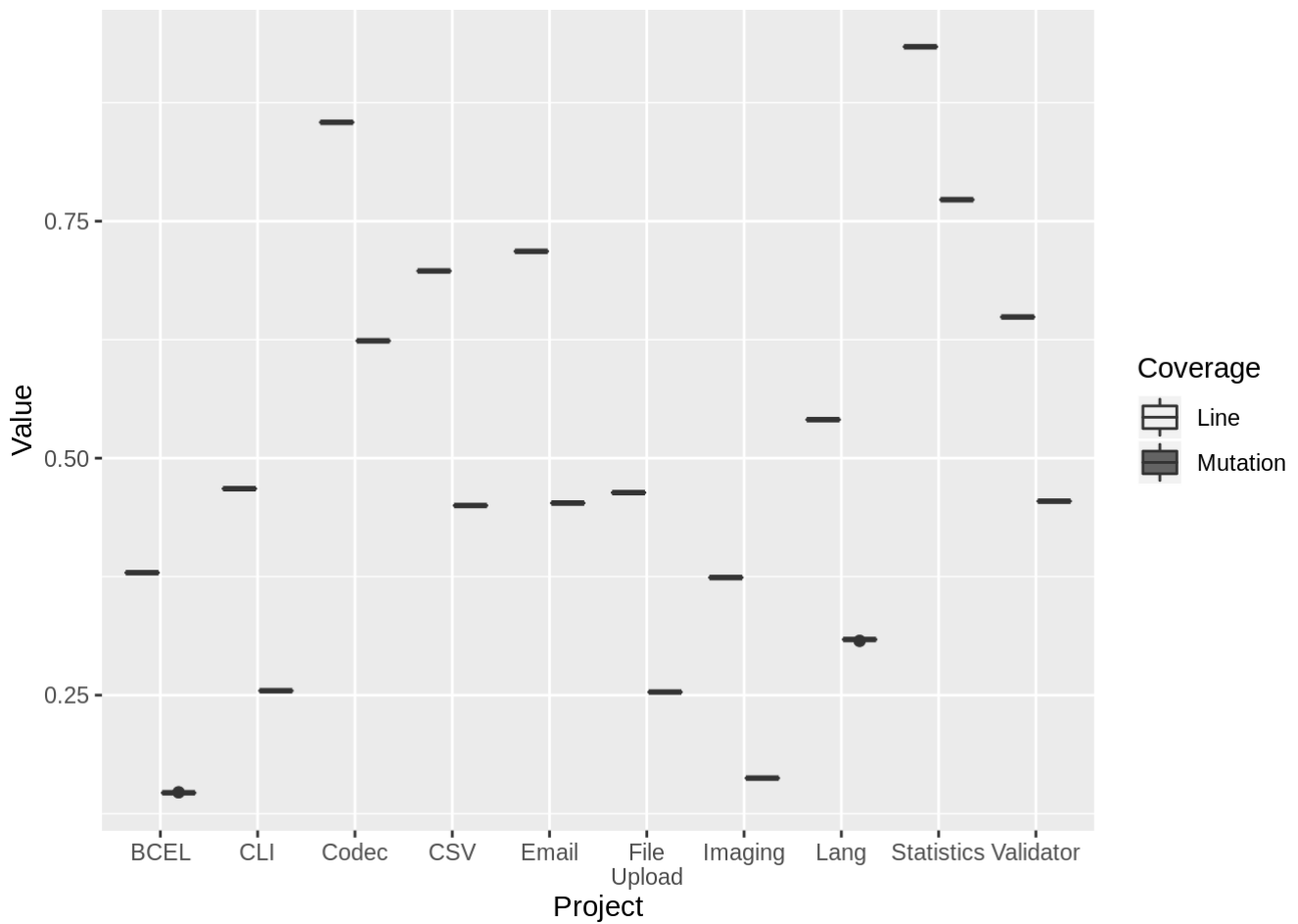
EvoSuite

```
ggplot(data = pitEvoMelted, aes(x=Project, y=value)) +
  geom_boxplot(aes(fill=variable)) +
  scale_x_discrete(labels=projects) +
  labs(x = "Project", y = "Value") +
  scale_fill_manual(values = c("#f0f0f0", "#636363"), name = "Coverage",
                    labels = c("Line", "Mutation"))
```



Randoop

```
ggplot(data = pitRandoopMelted, aes(x=Project, y=value)) +
  geom_boxplot(aes(fill=variable)) +
  scale_x_discrete(labels=projects) +
  labs(x = "Project", y = "Value") +
  scale_fill_manual(values = c("#f0f0f0", "#636363"), name = "Coverage",
                    labels = c("Line", "Mutation"))
```

Manual

```
ggplot(data = pitManualMelted, aes(x=Project, y=value)) +
  geom_boxplot(aes(fill=variable)) +
  scale_x_discrete(labels=projects) +
  labs(x = "Project", y = "Value") +
  scale_fill_manual(values = c("#f0f0f0", "#636363"), name = "Coverage",
                    labels = c("Line", "Mutation"))
```

