

Gender Research: RQ1 Analysis

(SETUP) Load the datasets

To answer RQ1, we use two datasets.

- all_committers_ds: contains information about all committers of the projects
- tf_ds: contains information about all key developers of the projects

```
all_committers_ds <- read.csv("../datasets/all_committers_repo.csv", head=T, sep=",")
tf_ds <- read.csv("../datasets/tf.csv", head=T, sep=";")
colnames(all_committers_ds)
```

```
## [1] "email"      "language"   "location"   "name"       "pais"       "gender"
## [7] "repo"
```

```
colnames(tf_ds)
```

```
## [1] "country"      "created_at"   "email"
## [4] "forks"        "full_name"    "gender"
## [7] "gender2"      "git_url"      "id"
## [10] "language"     "lines"        "location"
## [13] "login"        "num_contributors" "rate_commits"
## [16] "repository"   "size"         "user"
## [19] "user_num_commits" "watchers"
```

```
nrow(all_committers_ds)
```

```
## [1] 242621
```

```
nrow(tf_ds)
```

```
## [1] 2640
```

```
summary(tf_ds$gender)
```

```
##      female    male unisex
##      142      130   2311    57
```

```
summary(tf_ds$gender2)
```

```
## female    male
##      88    2552
```

```
tf_ds$num_contributors <- as.numeric(gsub(",", "", as.character(tf_ds$num_contributors)))
tf_ds$lines <- as.numeric(gsub(",", "", as.character(tf_ds$lines)))
tf_ds$size <- as.numeric(gsub(",", "", as.character(tf_ds$size)))
tf_ds$watchers <- as.numeric(gsub(",", "", as.character(tf_ds$watchers)))
```

Exploratory data analysis

- Number of distinct projects

```
## count(distinct full_name)
## 1                          1184
```

- Number of projects with at least 5 key developers

```
##          full_name totalTF
## 1      ansible/ansible      23
## 2      gitlabhq/gitlabhq      20
## 3      elastic/elasticsearch      17
## 4      kubernetes/kubernetes      17
## 5      python/cpython      15
## 6      apache/incubator-mxnet      13
## 7      facebook/react-native      13
## 8      pytorch/pytorch      13
## 9      rails/rails      13
## 10     jedi4ever/veewee      12
## 11     php/php-src      11
## 12     saltstack/salt      10
## 13     spree/spree      10
## 14     FFmpeg/FFmpeg      9
## 15     Microsoft/CNTK      9
## 16     aspnet/AspNetCore      9
## 17     dotnet/corefx      9
## 18     emberjs/ember.js      9
## 19     github/linguist      9
## 20     golang/go      9
## 21     moby/moby      9
## 22     Bash-it/bash-it      8
## 23     WordPress/WordPress      8
## 24     apache/kafka      8
## 25     chriskempson/tomorrow-theme      8
## 26     cockroachdb/cockroach      8
## 27     facebook/folly      8
## 28     geekcomputers/Python      8
## 29     mesosphere/marathon      8
## 30     puppetlabs/puppet      8
```

```
## [1] 88
```

```
## count(distinct full_name)
## 1      7.432432
```

- Characteristics of the projects

```
projects_ds <- sqldf("select full_name, lines, size, num_contributors, forks, watchers, count(distinct
                        from tf_ds
                        group by full_name, lines, size, num_contributors, forks, watchers)")

summary(projects_ds[,c("lines", "num_contributors", "size", "forks", "watchers")])
```

```
##      lines      num_contributors      size      forks
## Min.   :      0 Min.   : 0.0 Min.   :      9 Min.   :      8
## 1st Qu.:  5183 1st Qu.: 33.0 1st Qu.:  2648 1st Qu.:  531
## Median : 23690 Median : 82.0 Median : 10852 Median : 1077
## Mean   : 174414 Mean   : 199.9 Mean   :  78591 Mean   :  2337
## 3rd Qu.: 104933 3rd Qu.: 187.2 3rd Qu.:  45630 3rd Qu.:  2504
## Max.   :9442645 Max.   :8413.0 Max.   :8299557 Max.   :64712
##      NA's      :8
##      watchers
## Min.   : 1097
## 1st Qu.: 5315
```

```
## Median : 8003
## Mean   : 12206
## 3rd Qu.: 13472
## Max.   :300666
##
```

- It is better to remove smaller projects

```
tf_ds <- filter(tf_ds, lines >= 5183, tf_ds$num_contributors >= 33)
```

```
sqldf("select count(distinct full_name) from tf_ds")
```

```
## count(distinct full_name)
## 1 737
```

```
projects_ds <- sqldf("select full_name, lines, size, num_contributors, forks, watchers, count(distinct
                      from tf_ds
                      group by full_name, lines, size, num_contributors, forks, watchers)")
```

```
pds_summary <- as.data.frame(sapply(projects_ds[,c("lines", "num_contributors", "size", "forks", "watchers"),
print(xtable(t(pds_summary)), type="latex")
```

```
## % latex table generated in R 3.6.1 by xtable 1.8-4 package
## % Tue Jan 7 08:27:09 2020
## \begin{table}[ht]
## \centering
## \begin{tabular}{rrrrrrr}
## \hline
## & Min. & 1st Qu. & Median & Mean & 3rd Qu. & Max. \\
## \hline
## lines & 5191.00 & 19523.00 & 57013.00 & 259367.63 & 195265.00 & 9442645.00 \\
## num\_contributors & 33.00 & 80.00 & 145.00 & 292.77 & 297.00 & 8413.00 \\
## size & 368.00 & 6625.00 & 20503.00 & 94498.98 & 78353.00 & 3950679.00 \\
## forks & 54.00 & 774.00 & 1481.00 & 2949.94 & 3171.00 & 64712.00 \\
## watchers & 1145.00 & 5882.00 & 9039.00 & 14284.96 & 16418.00 & 300666.00 \\
## \hline
## \end{tabular}
## \end{table}
```

- Correlation: number of developers and number of TF developers
- Correlation: lines of code and number of TF developers

```
t3 <- sqldf("select full_name, num_contributors, lines, count(*) ignore
            from tf_ds
            group by full_name, num_contributors, lines
            order by 3 desc")
```

```
nrow(t3)
```

```
## [1] 737
```

```
head(t3)
```

```
##          full_name num_contributors  lines ignore
## 1  apple/turicreate             51 9442645      2
## 2   dotnet/coreclr             556 9290723      2
## 3   nodejs/node             2424 5421767      7
```

```
## 4      mongodb/mongo      365 4355800      7
## 5      dotnet/roslyn      349 4103895      3
## 6  kubernetes/kubernetes  2092 3566770     19

t4 <- merge(t2, t3)
cor.test(as.numeric(t4$num_contributors), t4$totalTF, method="spearman")

## Warning in cor.test.default(as.numeric(t4$num_contributors), t4$totalTF, :
## Cannot compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  as.numeric(t4$num_contributors) and t4$totalTF
## S = 61991, p-value = 7.052e-05
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.4151481
```

```
cor.test(as.numeric(t4$lines), t4$totalTF, method="spearman")
```

```
## Warning in cor.test.default(as.numeric(t4$lines), t4$totalTF, method =
## "spearman"): Cannot compute exact p-value with ties

##
## Spearman's rank correlation rho
##
## data:  as.numeric(t4$lines) and t4$totalTF
## S = 80164, p-value = 0.02375
## alternative hypothesis: true rho is not equal to 0
## sample estimates:
##      rho
## 0.243699
```

- Total number of key developers

```
distinct_tfs <- sqldf("select full_name, count(distinct login) total
                      from tf_ds
                      group by full_name
                      order by 2 desc")
```

```
nrow(distinct_tfs)
```

```
## [1] 737
```

```
head(distinct_tfs, 50)
```

```
##           full_name total
## 1      ansible/ansible    23
## 2      gitlabhq/gitlabhq    20
## 3      elastic/elasticsearch    17
## 4      kubernetes/kubernetes    17
## 5      python/cpython      15
## 6  apache/incubator-mxnet    13
## 7      facebook/react-native    13
## 8      pytorch/pytorch      13
## 9      rails/rails         13
```

```
## 10          jedi4ever/veewee      12
## 11          php/php-src          11
## 12          spree/spree          10
## 13          FFmpeg/FFmpeg        9
## 14          Microsoft/CNTK        9
## 15          aspnet/AspNetCore     9
## 16          dotnet/corefx         9
## 17          emberjs/ember.js      9
## 18          github/linguist       9
## 19          golang/go             9
## 20          moby/moby             9
## 21          saltstack/salt        9
## 22          Bash-it/bash-it       8
## 23          WordPress/WordPress   8
## 24          apache/kafka          8
## 25          chriskempson/tomorrow-theme 8
## 26          cockroachdb/cockroach 8
## 27          facebook/folly        8
## 28          mesosphere/marathon    8
## 29          puppetlabs/puppet     8
## 30          robbyrussell/oh-my-zsh 8
## 31          twitter/finagle        8
## 32          RaRe-Technologies/gensim 7
## 33          fzaninotto/Faker       7
## 34          geekcomputers/Python   7
## 35          influxdata/influxdb    7
## 36          nodejs/node            7
## 37          opencv/opencv          7
## 38          palantir/tslint         7
## 39          rubocop-hq/rubocop      7
## 40          twitter/scalding        7
## 41 windows-toolkit/WindowsCommunityToolkit 7
## 42          Microsoft/vscode        6
## 43          PaddlePaddle/Paddle     6
## 44          Theano/Theano            6
## 45          akka/akka               6
## 46          angular/angular         6
## 47          apache/thrift           6
## 48          apple/swift             6
## 49          chef/chef              6
## 50          dotnet/cli              6
```

```
nrow(distinct_tfs[distinct_tfs$total>1, ])
```

```
## [1] 397
```

(RQ1) How common are women key developers in OSS projects?

We answer this research question using an exploratory data analysis. We first report the characteristics of the projects (see table below).

```
ds_summary <- sqldf("select language, full_name, lines, num_contributors, forks, watchers, count(distinct
                      from tf_ds
                      group by language, full_name, lines, num_contributors, forks, watchers)")
```

```
nrow(ds_summary)
```

```
[1] 737
```

```
ds_summary_language <- sqldf("select language as 'Prog. Language',  
                             avg(lines) as 'Average number of lines of code',  
                             avg(num_contributors) as 'Average number of contributors',  
                             avg(forks) as 'Average number of forks' ,  
                             avg(watchers) as 'Average number of watchers',  
                             avg(num_tf) as 'Average number of key developers'  
                             from ds_summary  
                             group by language  
                             order by 1")  
  
print(xtable(ds_summary_language), type="html")
```

Prog. Language

Average number of lines of code

Average number of contributors

Average number of forks

Average number of watchers

Average number of key developers

1

C

241308.02

232.54

2350.66

9825.80

2.00

2

C#

453339.08

161.11

1656.71

5909.29

2.24

3

C++

633316.91

253.04

3323.22
13645.39
3.01
4
CSS
120447.17
108.91
2313.78
13286.96
1.57
5
Go
489145.35
282.82
2520.96
15314.19
2.78
6
Java
275866.11
217.61
5640.75
16677.05
2.43
7
JavaScript
202729.63
505.96
7611.28
42959.09
2.54
8
Objective-C
165398.92
102.00
1650.27

8826.46
1.62
9
PHP
108836.46
289.46
2007.04
9518.15
1.87
10
Python
162283.81
493.00
4397.00
18717.02
3.29
11
Ruby
99639.68
601.07
2261.10
10418.13
3.10
12
Scala
100119.61
157.12
978.74
3646.35
2.42
13
Shell
66032.21
225.71
1843.12
12075.75

2.88

14

Swift

33857.72

114.69

1347.94

10810.28

1.69

15

TypeScript

245799.41

321.70

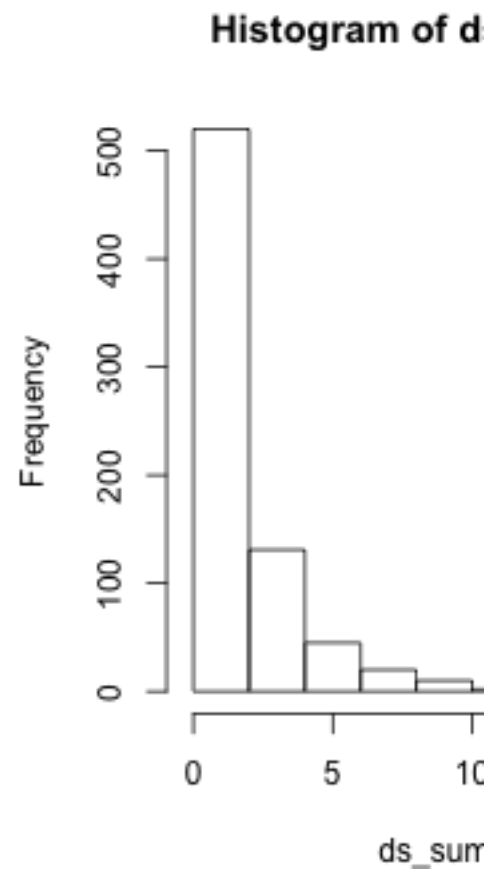
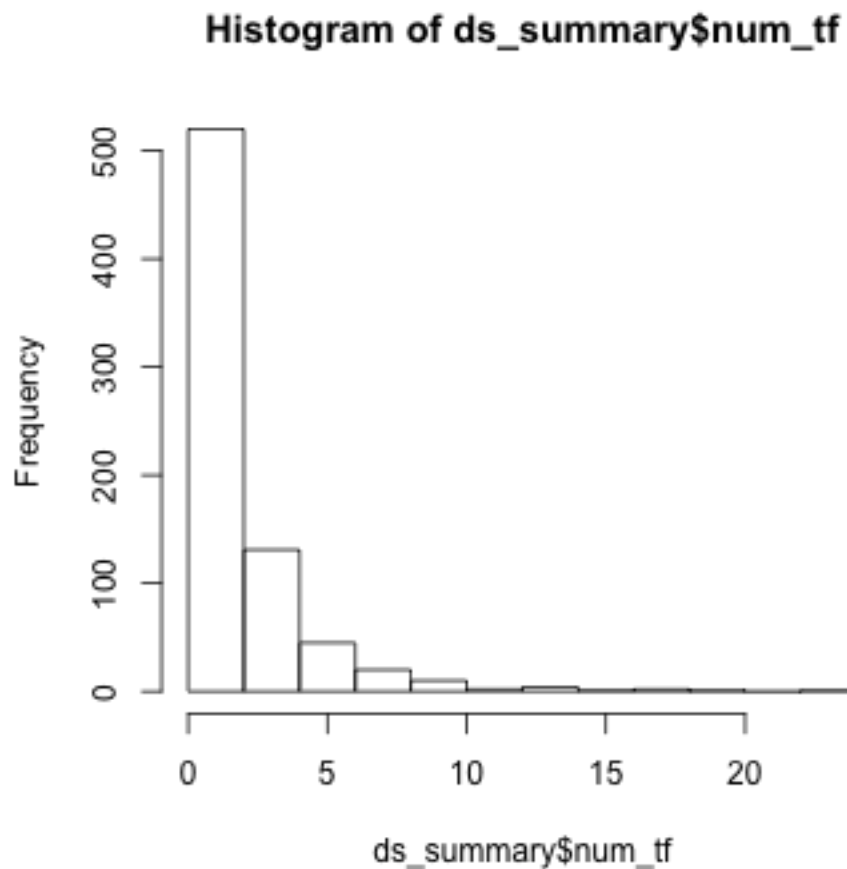
2228.66

12903.53

2.00

Next we present a histogram with the number of key developers per project.

```
plot(hist(ds_summary$num_tf))
```



Let's try to understand the women participation on OSS projects.

```
all_committers_ds["gender_final"] <-
  ifelse(trim(as.character(all_committers_ds$gender)) == "", as.character("unisex"), as.character(all_committers_ds$gender))

tf_ds["gender_final"] <-
  ifelse(as.character(tf_ds$gender) != as.character(tf_ds$gender2), "unisex", as.character(tf_ds$gender))

# note. we have to remove duplicated data

t_all_committers <- sqldf("select name, gender_final, count(*) summaryCommitters
                           from all_committers_ds
                           group by name, gender_final
                           order by 3 desc")

t_tf <- sqldf("select login, user, gender_final, count(*) summaryTF
               from tf_ds
               group by login, user, gender_final
               order by 3 desc")
```

```
sqldf("select gender_final, count(*) total from all_committers_ds group by gender_final")
```

```
##  gender_final  total
## 1      female 12987
## 2      male 208384
## 3      unisex  21250
```

```
sqldf("select gender_final, count(*) total from tf_ds group by gender_final")
```

```
##  gender_final  total
## 1      female    45
## 2      male  1762
## 3      unisex   195
```

```
slices <- c(21250, 12987, 208384)
lbls <- c("unisex", "female", "male")
pct <- round(slices/sum(slices)*100)
lbls <- paste(lbls, pct) # add percents to labels
lbls <- paste(lbls,"%",sep="") # ad % to labels
```

```
pie(slices, labels=lbls,explode=0.1,main="Pie Chart of Contributors")
```

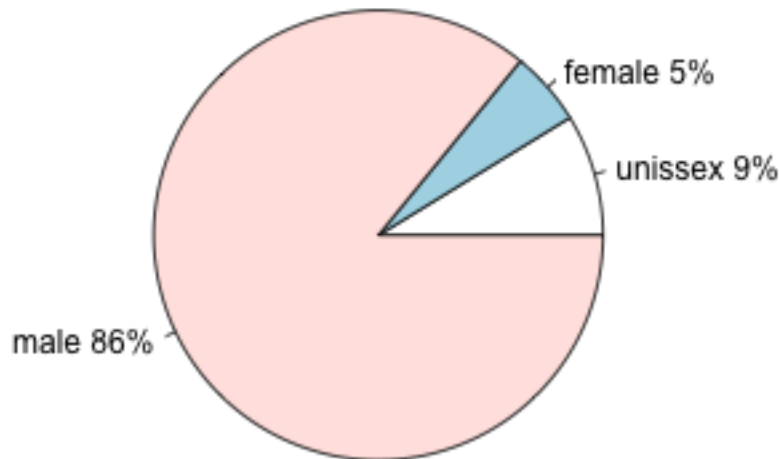
```
## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "explode" is not a graphical parameter
```

```
## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "explode" is not a graphical parameter
```

```
## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "explode" is not a graphical parameter
```

```
## Warning in title(main = main, ...): "explode" is not a graphical parameter
```

Pie Chart of Contributors



```
slices <- c(195, 45, 1762)
lbls <- c("unissex", "female", "male")
pct <- slices/sum(slices)*100
lbls <- paste(lbls, pct) # add percents to labels
lbls <- paste(lbls,"%",sep="") # ad % to labels

pie(slices, labels=lbls,explode=0.1,main="Key developers")

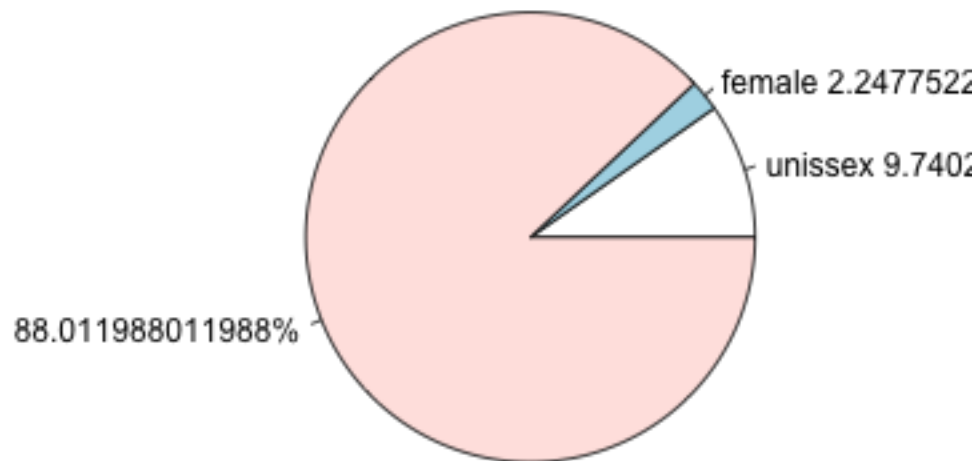
## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "explode" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "explode" is not a graphical parameter

## Warning in text.default(1.1 * P$x, 1.1 * P$y, labels[i], xpd = TRUE, adj =
## ifelse(P$x < : "explode" is not a graphical parameter

## Warning in title(main = main, ...): "explode" is not a graphical parameter
```

Key developers



That is, the percentage of women key developers is smaller than the percentage of women, when considering all contributors of the projects.

Now, let's group the key developers by language and gender.

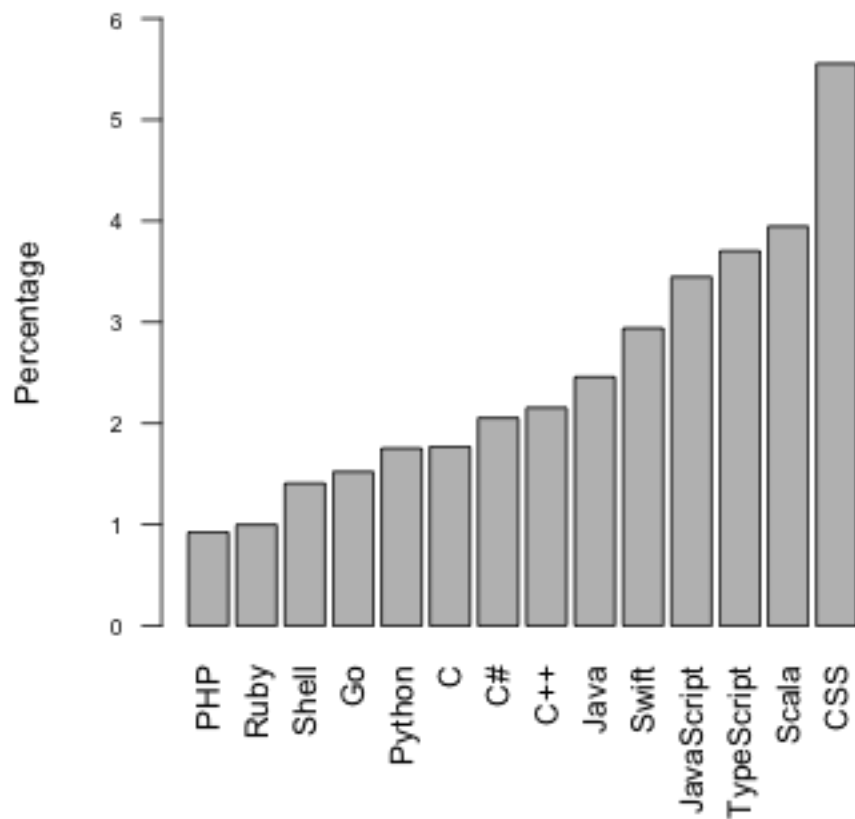
```
tf_language_female_only <- sqldf("select language, count(*) total_female
                                   from tf_ds
                                   where gender_final = 'female'
                                   group by language");
```

```
tf_language <- sqldf("select language, count(*) total
                      from tf_ds
                      group by language")
```

```
res <- merge(tf_language, tf_language_female_only)
res["percent"] <- res$total_female * 100 / res$total
```

```
res <- res[order(res$percent), ]
```

```
barplot(res$percent, names.arg = res$language, las=2, ylab="Percentage", cex.axis = 0.7, ylim=range(pre
```



```
# dotplot(reorder(res$language, res$percent)~res$language)

# dotplot(res$percent~res$language, las=2)

# dotchart2(res$percent, labels=res$language, las=2, horizontal=F, sort.=T)
```

How many projects have at least one women key developer?

```
sqldf("select count(distinct full_name) total_projects
      from tf_ds
      where gender_final = 'female');"
```

```
## total_projects
## 1              37
```

```
sqldf("select count(distinct full_name) total_projects
      from tf_ds");
```

```
## total_projects
## 1              737
```

Table 1.

```
head(ds_summary, results='asis')
```

```
## language                                full_name  lines
```

## 1	C	FFmpeg/FFmpeg	1179881
## 2	C	MarlinFirmware/Marlin	117328
## 3	C	SamyPesse/How-to-Make-a-Computer-Operating-System	23443
## 4	C	alibaba/tengine	279209
## 5	C	allinurl/goaccess	26906
## 6	C	beanstalkd/beanstalkd	7154
##	num_contributors	forks	watchers
## 1	1019	5424	14444
## 2	413	6910	5211
## 3	35	3127	17933
## 4	67	1985	8229
## 5	75	624	8676
## 6	45	734	5005

```
tab1<- sqldf("select language, count(distinct full_name) projects, sum(num_contributors) contributors
from ds_summary
group by language")
```

```
tab1_kd <- sqldf("select language, count(*) as total_kds
from tf_ds
group by language")
```

```
tab1_men <- sqldf("select language, count(*) as total_male
from tf_ds
where gender_final = 'male'
group by language")
```

```
tab1_female <- sqldf("select language, count(*) as total_female
from tf_ds
where gender_final = 'female'
group by language")
```

```
tab1_unknown <- sqldf("select language, count(*) as total_unknown
from tf_ds
where gender_final = 'unisex'
group by language")
```

```
tab1_total_projects_with_women <- sqldf("select language, count(distinct full_name) as total_projects_w
from tf_ds
where gender_final = 'female'
group by language")
```

```
tab1 <- merge(tab1, tab1_kd)
```

```
tab1 <- merge(tab1, tab1_men)
```

```
tab1 <- merge(tab1, tab1_female)
```

```
tab1 <- merge(tab1, tab1_unknown)
```

```
tab1 <- merge(tab1, tab1_total_projects_with_women)
```

```

tab1["percentage"] <- tab1$total_projects_with_women * 100 / tab1$projects

print(xtable(tab1[,c("language", "projects", "contributors", "total_kds", "total_male", "total_female",

## % latex table generated in R 3.6.1 by xtable 1.8-4 package
## % Tue Jan 7 08:27:14 2020
## \begin{table}[ht]
## \centering
## \begin{tabular}{rlrrrrrrrr}
## \hline
## & language & projects & contributors & total\_kds & total\_male & total\_female & total\_unknown &
## \hline
## 1 & C & 50 & 11627.00 & 113 & 99 & 2 & 12 & 2 & 4.00 \\
## 2 & C\# & 63 & 10150.00 & 146 & 129 & 3 & 14 & 3 & 4.76 \\
## 3 & C++ & 67 & 16954.00 & 232 & 191 & 5 & 36 & 3 & 4.48 \\
## 4 & CSS & 23 & 2505.00 & 36 & 30 & 2 & 4 & 2 & 8.70 \\
## 5 & Go & 68 & 19232.00 & 197 & 169 & 3 & 25 & 3 & 4.41 \\
## 6 & Java & 44 & 9575.00 & 122 & 109 & 3 & 10 & 2 & 4.55 \\
## 7 & JavaScript & 67 & 33899.00 & 203 & 172 & 7 & 24 & 5 & 7.46 \\
## 8 & PHP & 46 & 13315.00 & 108 & 98 & 1 & 9 & 1 & 2.17 \\
## 9 & Python & 42 & 20706.00 & 171 & 155 & 3 & 13 & 2 & 4.76 \\
## 10 & Ruby & 60 & 36064.00 & 200 & 183 & 2 & 15 & 2 & 3.33 \\
## 11 & Scala & 57 & 8956.00 & 152 & 137 & 6 & 9 & 5 & 8.77 \\
## 12 & Shell & 24 & 5417.00 & 71 & 63 & 1 & 7 & 1 & 4.17 \\
## 13 & Swift & 36 & 4129.00 & 68 & 59 & 2 & 7 & 2 & 5.56 \\
## 14 & TypeScript & 64 & 20589.00 & 135 & 123 & 5 & 7 & 4 & 6.25 \\
## \hline
## \end{tabular}
## \end{table}

sum(tab1$projects)

## [1] 711

sum(tab1$contributors)

## [1] 213118

sum(tab1$total_kds)

## [1] 1954

sum(tab1$total_male)

## [1] 1717

sum(tab1$total_female)

## [1] 45

sum(tab1$total_unknown)

## [1] 192

sum(tab1$total_projects_with_women)

## [1] 37

```



```
mean(tab1$percentage)
```

```
## [1] 5.240598
```

```
sd(tab1$percentage)
```

```
## [1] 1.929702
```