# Dataset Processing Code. R

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```
# Create Census pay train set, and validation set
# Note: this process could take a couple of minutes
if (!require(tidyverse))
 install.packages("tidyverse", repos = "http://cran.us.r-project.org")
## Loading required package: tidyverse
## -- Attaching packages ------ tidyverse 1.3.1 --
## v ggplot2 3.3.5 v purrr 0.3.4
## v tibble 3.1.2 v dplyr 1.0.7
## v tidyr 1.1.3 v stringr 1.4.0
## v readr 1.4.0 v forcats 0.5.1
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
if (!require(caret))
 install.packages("caret", repos = "http://cran.us.r-project.org")
## Loading required package: caret
## Loading required package: lattice
## Attaching package: 'caret'
## The following object is masked from 'package:purrr':
##
      lift
##
if (!require(data.table))
 install.packages("data.table", repos = "http://cran.us.r-project.org")
```

```
## Loading required package: data.table
##
## Attaching package: 'data.table'
## The following objects are masked from 'package:dplyr':
##
##
       between, first, last
## The following object is masked from 'package:purrr':
##
##
       transpose
if (!require(dplyr))
  install.packages("dplyr", repos = "http://cran.us.r-project.org")
if (!require(gridExtra))
  install.packages("gridExtra", repos = "http://cran.us.r-project.org")
## Loading required package: gridExtra
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
##
       combine
if (!require(kableExtra))
  install.packages("kableExtra", repos = "http://cran.us.r-project.org")
## Loading required package: kableExtra
##
## Attaching package: 'kableExtra'
## The following object is masked from 'package:dplyr':
##
##
       group_rows
if (!require(epiDisplay))
  install.packages("epiDisplay")
## Loading required package: epiDisplay
## Loading required package: foreign
## Loading required package: survival
```

```
##
## Attaching package: 'survival'
## The following object is masked from 'package:caret':
##
##
       cluster
## Loading required package: MASS
##
## Attaching package: 'MASS'
## The following object is masked from 'package:dplyr':
##
##
       select
## Loading required package: nnet
##
## Attaching package: 'epiDisplay'
## The following object is masked from 'package:lattice':
##
##
       dotplot
## The following object is masked from 'package:ggplot2':
##
##
       alpha
library(tidyverse)
library(caret)
library(data.table)
library(dplyr)
library(gridExtra)
library(kableExtra)
library(epiDisplay)
# Adult Census Income
# https://www.kaggle.com/uciml/adult-census-income
#download the dataset from the staging github location
dl <- tempfile()</pre>
download.file("https://github.com/rajeshharidas/havardxwork2/raw/main/adult.csv.zip",
              d1)
#read all the data into R dataset
adultpay <-
  fread(
    text = gsub(",", "\t", readLines(unzip(dl, "adult.csv"))),
    col.names = c(
      "age",
```

```
"workclass",
      "fnlwgt",
      "education",
      "education.num",
      "marital.status",
      "occupation",
      "relationship",
      "race",
      "sex",
     "capital.gain",
      "capital.loss",
      "hours.per.week",
      "native.country",
      "income"
   )
  )
#Keep only USA data
#Remove '?' from the work class and rename it to class, and finally remove workclass
#Rename all columns with a '.' in it
#Remove capital gain and loss column
#remove non-alphanumeric character from column data
#rename the label for below and above 50K income
adultpayclean <-
  adultpay %>% filter (native.country == 'United-States') %>%
 mutate (class = ifelse(workclass == '?', 'Unknown', str_replace_all(workclass, "-", ""))) %>%
 dplyr::select(-workclass, -capital.gain, -capital.loss) %>%
 rename(
   с(
     eduyears = education.num,
     maritalstatus = marital.status,
     hoursperweek = hours.per.week,
     native = native.country
   )
 ) %>%
  mutate (maritalstatus = ifelse(
   maritalstatus == '?',
   'Unknown',
   str_replace_all(maritalstatus, "-", "")
  )) %>%
  mutate (occupation = ifelse(
   occupation == '?',
   'Unknown',
   str_replace_all(occupation, "-", "")
  )) %>%
  mutate (education = ifelse(education == '?', 'Unknown', str_replace_all(education, "-", ""))) %>%
  mutate (relationship = ifelse(
   relationship == '?',
   'Unknown',
   str_replace_all(relationship, "-", "")
 )) %>%
  mutate (native = ifelse(native == '?', 'Unknown', str_replace_all(native, "-", ""))) %>%
  mutate (income = ifelse(
```

```
income == '?',
    'Unknown',
    str_replace_all(income, "<=50K", "AtBelow50K")</pre>
  mutate (income = ifelse(
    income == '?',
    'Unknown',
   str_replace_all(income, ">50K", "Above50K")
  ))
# R 4.0 or later:
#convert all the character labels to factors
adultpayclean <-
  as.data.frame(adultpayclean) %>% mutate(
   education = as.factor(education),
   maritalstatus = as.factor(maritalstatus),
   occupation = as.factor(occupation),
   relationship = as.factor(relationship),
   race = as.factor(race),
   sex = as.factor(sex),
   class = as.factor(class),
   income = as.factor(income)
  )
# Validation set will be 10% of adultpay data
set.seed(1, sample.kind = "Rounding") # if using R 3.5 or earlier, use `set.seed(1)`
## Warning in set.seed(1, sample.kind = "Rounding"): non-uniform 'Rounding' sampler
## used
test index <-
  createDataPartition(
   y = adultpayclean$income,
   times = 1,
   p = 0.1,
   list = FALSE
  )
adultpayclean_train <- adultpayclean[-test_index, ]</pre>
adultpayclean_validation <- adultpayclean[test_index, ]</pre>
glimpse(adultpay)
## Rows: 32,561
## Columns: 15
## $ age
                    <int> 90, 82, 66, 54, 41, 34, 38, 74, 68, 41, 45, 38, 52, 32,~
                    <chr> "?", "Private", "?", "Private", "Private", "Private", "~
## $ workclass
## $ fnlwgt
                    <int> 77053, 132870, 186061, 140359, 264663, 216864, 150601, ~
                    <chr> "HS-grad", "HS-grad", "Some-college", "7th-8th", "Some-~
## $ education
## $ education.num <int> 9, 9, 10, 4, 10, 9, 6, 16, 9, 10, 16, 15, 13, 14, 16, 1~
## $ marital.status <chr> "Widowed", "Widowed", "Widowed", "Divorced", "Separated~
## $ occupation
                    <chr> "?", "Exec-managerial", "?", "Machine-op-inspct", "Prof~
## $ relationship <chr> "Not-in-family", "Not-in-family", "Unmarried", "Unmarri~
```

<int> 90, 82, 66, 54, 41, 34, 38, 74, 68, 45, 38, 52, 32, 51, ~

<int> 77053, 132870, 186061, 140359, 264663, 216864, 150601, 8~

dim(adultpayclean)

## \$ age

## \$ fnlwgt

**##** [1] 29170 13

dim(adultpayclean\_train)

## [1] 26252 13

dim(adultpayclean\_validation)

## [1] 2918 13

summary(adultpayclean)

```
##
                        fnlwgt
                                            education
                                                            eduyears
         age
                                                               : 1.00
##
  Min.
          :17.00
                   Min.
                         : 12285
                                      HSgrad
                                                 :9702
                                                         Min.
   1st Qu.:28.00
                   1st Qu.: 115895
                                      Somecollege:6740
                                                         1st Qu.: 9.00
##
  Median :37.00
                   Median: 176730
                                      Bachelors :4766
                                                         Median :10.00
## Mean
          :38.66
                    Mean : 187069
                                      Masters
                                                 :1527
                                                         Mean
                                                               :10.17
                    3rd Qu.: 234139
## 3rd Qu.:48.00
                                      Assocvoc
                                                 :1289
                                                         3rd Qu.:12.00
## Max.
          :90.00
                   Max.
                          :1484705
                                      11th
                                                 :1067
                                                         Max.
                                                                :16.00
##
                                      (Other)
                                                 :4079
##
               maritalstatus
                                         occupation
                                                             relationship
##
                               Execmanagerial:3735 Husband
                                                                   :11861
                       : 4162
  Divorced
```

Profspecialty :3693 MarriedAFspouse 23 Notinfamily: 7528 ## Marriedcivspouse :13368 Craftrepair :3685 Otherrelative: 696 Marriedspouseabsent: Admclerical Ownchild : 4691 ## 253 :3449 Nevermarried : 9579 Sales :3364 Unmarried : 3033 ## Separated ## 883 Otherservice :2777 Wife : 1361 ## Widowed 902 (Other) :8467 ## sex hoursperweek native race ## Amer-Indian-Eskimo: 296 Female: 9682 : 1.00 Length: 29170 Min. ## Asian-Pac-Islander: 292 Male :19488 1st Qu.:40.00 Class : character ## Black : 2832 Median :40.00 Mode :character ## Other 129 Mean :40.45 ## White 3rd Qu.:45.00 :25621 :99.00 ## Max. ## ## income class ## Above50K : 7171 Private :20135 ## AtBelow50K:21999 Selfempnotinc: 2313 : 1956 ## Localgov ## Unknown : 1659 ## Stategov : 1210 ## Selfempinc 991 ## (Other) 906

#### summary(adultpayclean\_train)

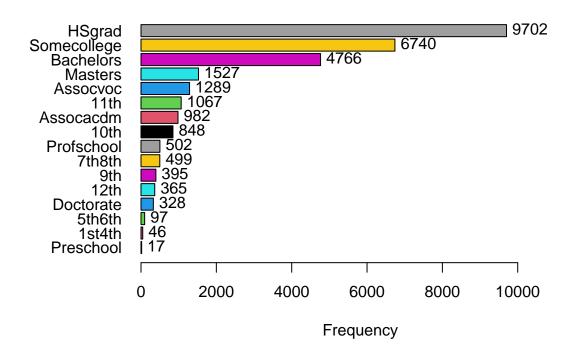
## fnlwgt education eduyears age ## Min. :17.00 Min. : 12285 **HSgrad** :8716 Min. : 1.00 1st Qu.:28.00 1st Qu.: 116052 Somecollege:6071 1st Qu.: 9.00 ## ## Median :37.00 Median : 176904 Bachelors :4318 Median :10.00 :38.66 ## Mean Mean : 187117 Masters :1366 Mean :10.17 3rd Qu.: 234099 ## 3rd Qu.:48.00 Assocvoc :1164 3rd Qu.:12.00 ## Max. :90.00 Max. :1484705 11th : 948 Max. :16.00 (Other) ## :3669 ## maritalstatus occupation relationship Divorced : 3757 Execmanagerial:3382 ## Husband :10674 ## MarriedAFspouse 21 Profspecialty:3318 Notinfamily : 6803 ## Marriedcivspouse Craftrepair :3300 Otherrelative: 629 :12033 Marriedspouseabsent: 229 Admclerical :3095 Ownchild : 4213 : 8616 ## Nevermarried Sales :3035 Unmarried : 2707 Separated 792 Otherservice :2490 ## Wife : 1226 Widowed 804 (Other) :7632 ## ## race sex hoursperweek native ## 261 Amer-Indian-Eskimo: Female: 8708 Min. : 1.00 Length: 26252 265 1st Qu.:40.00 ## Asian-Pac-Islander: Male :17544 Class : character ## Black : 2537 Median :40.00 Mode :character ## Other 119 Mean :40.47 ## White :23070 3rd Qu.:45.00 ## Max. :99.00 ## ## income class ## Above50K : 6453 Private :18093 ## AtBelow50K:19799 Selfempnotinc: 2087 ## Localgov : 1777 Unknown : 1494 ##

Dataset	Number of Rows	Number of Columns
train	26252	13
validation	2918	13

```
## Stategov : 1081
## Selfempinc : 910
## (Other) : 810
```

```
tab1(adultpayclean$education, sort.group = "decreasing", cum.percent = TRUE)
```

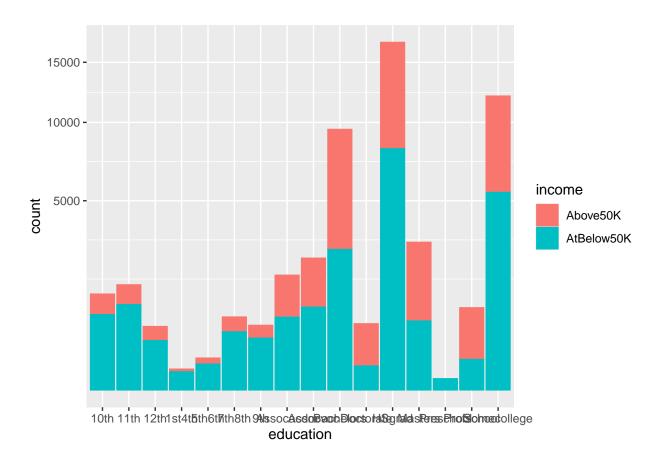
### Distribution of adultpayclean\$education



```
## adultpayclean$education :
##
               Frequency Percent Cum. percent
## HSgrad
                    9702
                            33.3
                                         33.3
## Somecollege
                    6740
                            23.1
                                         56.4
## Bachelors
                    4766
                            16.3
                                         72.7
                    1527
                            5.2
                                         77.9
## Masters
```

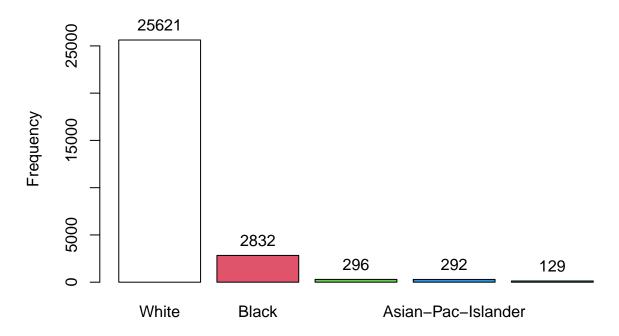
```
1289
                             4.4
                                         82.4
## Assocvoc
                    1067
                             3.7
                                         86.0
## 11th
                                         89.4
## Assocacdm
                    982
                             3.4
## 10th
                     848
                             2.9
                                         92.3
## Profschool
                     502
                             1.7
                                         94.0
## 7th8th
                     499
                             1.7
                                         95.7
## 9th
                     395
                             1.4
                                         97.1
                             1.3
                                         98.3
## 12th
                     365
## Doctorate
                     328
                             1.1
                                         99.5
## 5th6th
                      97
                             0.3
                                         99.8
## 1st4th
                      46
                             0.2
                                         99.9
                      17
                             0.1
                                        100.0
## Preschool
    Total
                   29170
                           100.0
                                        100.0
```

```
adultpayclean %>% group_by(education) %>%
mutate(n=n()) %>% ggplot() +
geom_bar(aes(education,col=income,fill=income)) + scale_y_sqrt()
```



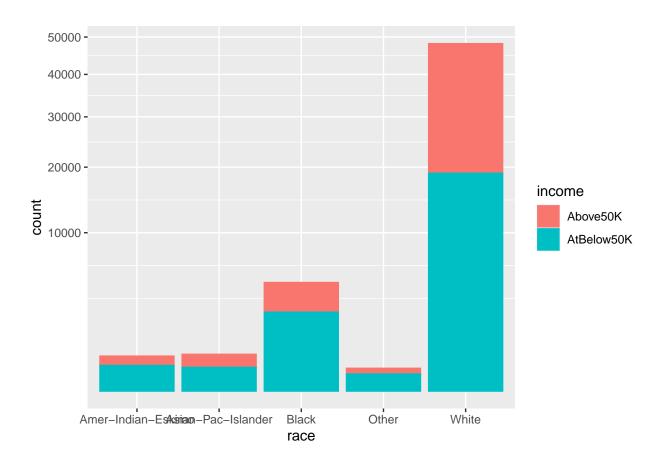
tab1(adultpayclean\$race, sort.group = "decreasing", cum.percent = TRUE)

## Distribution of adultpayclean\$race



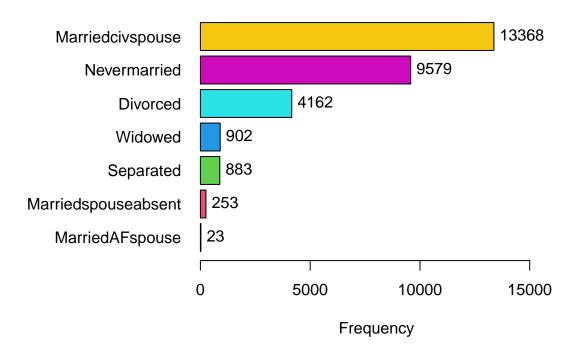
```
## adultpayclean$race :
                      Frequency Percent Cum. percent
## White
                           25621
                                    87.8
                                                 87.8
## Black
                            2832
                                     9.7
                                                 97.5
                                                 98.6
## Amer-Indian-Eskimo
                             296
                                     1.0
## Asian-Pac-Islander
                                                 99.6
                             292
                                     1.0
## Other
                             129
                                     0.4
                                                100.0
     Total
                           29170
                                   100.0
                                                100.0
```

```
adultpayclean %>% group_by(race) %>%
mutate(n=n()) %>% ggplot() +
geom_bar(aes(race,col=income,fill=income)) + scale_y_sqrt()
```

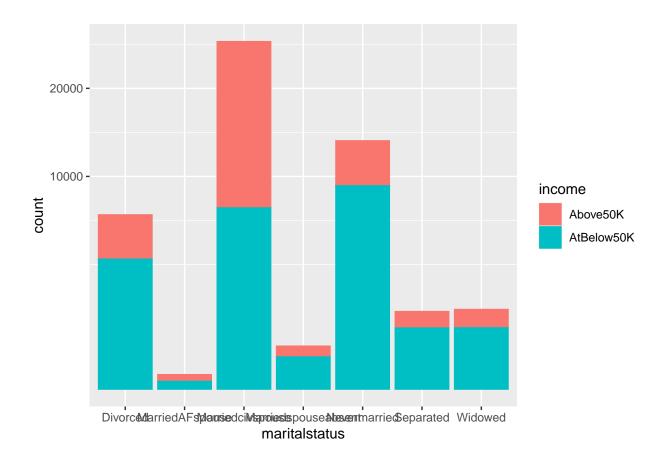


tab1(adultpayclean\$maritalstatus, sort.group = "decreasing", cum.percent = TRUE)

### Distribution of adultpayclean\$maritalstatus



```
## adultpayclean$maritalstatus :
                       Frequency Percent Cum. percent
## Marriedcivspouse
                           13368
                                     45.8
                                                  45.8
                                     32.8
                                                  78.7
## Nevermarried
                            9579
## Divorced
                            4162
                                     14.3
                                                  92.9
## Widowed
                             902
                                      3.1
                                                  96.0
## Separated
                             883
                                      3.0
                                                  99.1
## Marriedspouseabsent
                              253
                                      0.9
                                                  99.9
## MarriedAFspouse
                              23
                                      0.1
                                                 100.0
     Total
                           29170
                                   100.0
                                                 100.0
adultpayclean %>% group_by(maritalstatus) %>%
  mutate(n=n()) %>% ggplot() +
  geom_bar(aes(maritalstatus,col=income,fill=income)) + scale_y_sqrt()
```

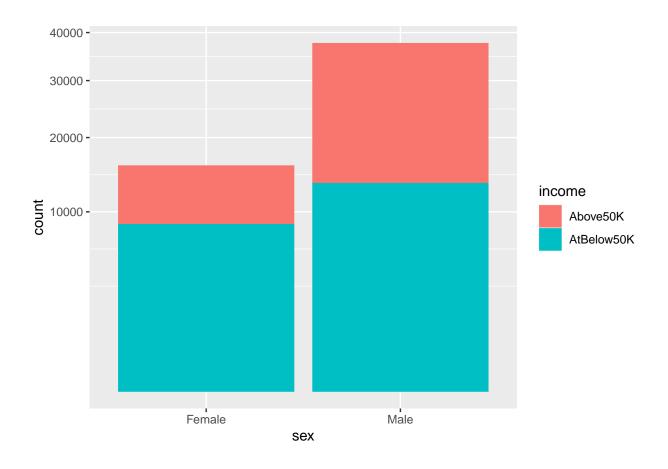


tab1(adultpayclean\$sex, sort.group = "decreasing", cum.percent = TRUE)

# Distribution of adultpayclean\$sex

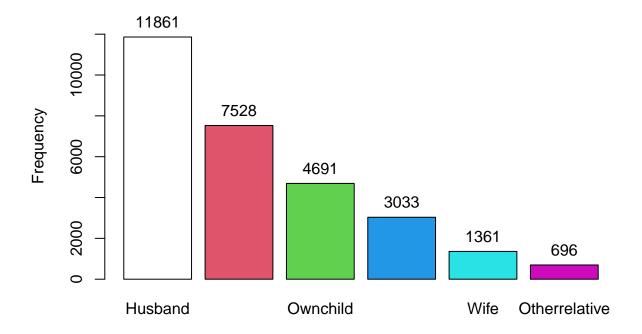


```
## adultpayclean$sex :
          Frequency Percent Cum. percent
## Male
              19488
                        66.8
                                    66.8
## Female
               9682
                        33.2
                                    100.0
              29170
                       100.0
                                    100.0
##
     Total
adultpayclean %>% group_by(sex) %>%
 mutate(n=n()) %>% ggplot() +
  geom_bar(aes(sex,col=income,fill=income)) + scale_y_sqrt()
```



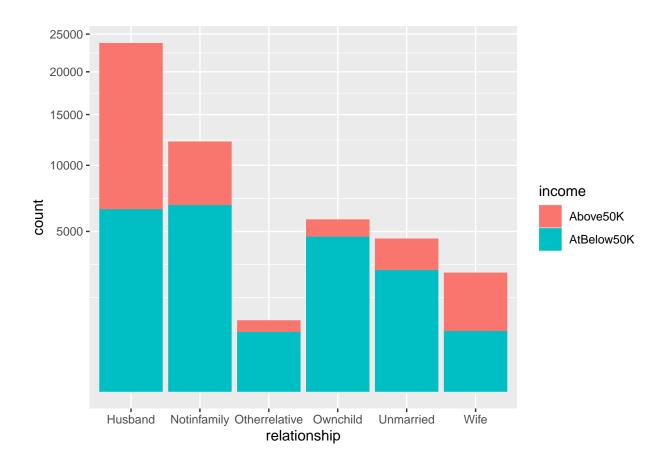
tab1(adultpayclean\$relationship, sort.group = "decreasing", cum.percent = TRUE)

## Distribution of adultpayclean\$relationship



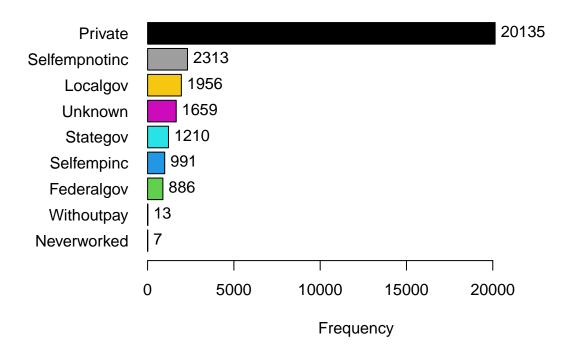
```
## adultpayclean$relationship :
                 Frequency Percent Cum. percent
## Husband
                     11861
                              40.7
                                            40.7
## Notinfamily
                      7528
                              25.8
                                            66.5
## Ownchild
                      4691
                              16.1
                                            82.6
## Unmarried
                      3033
                              10.4
                                            92.9
## Wife
                      1361
                               4.7
                                            97.6
## Otherrelative
                       696
                               2.4
                                           100.0
     Total
                     29170
                             100.0
                                           100.0
```

```
adultpayclean %>% group_by(relationship) %>%
mutate(n=n()) %>% ggplot() +
geom_bar(aes(relationship,col=income,fill=income)) + scale_y_sqrt()
```

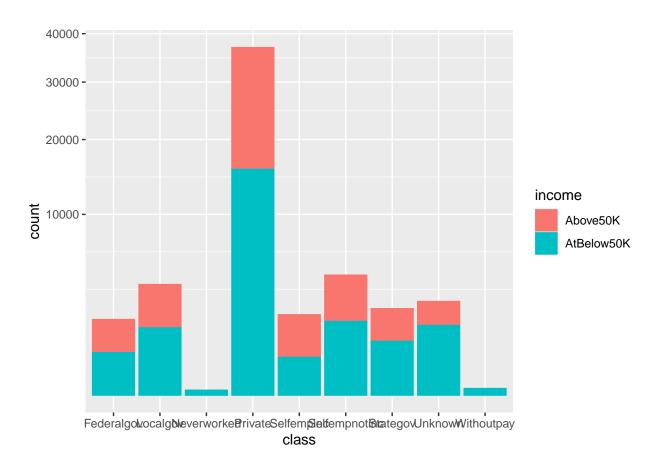


tab1(adultpayclean\$class, sort.group = "decreasing", cum.percent = TRUE)

### Distribution of adultpayclean\$class

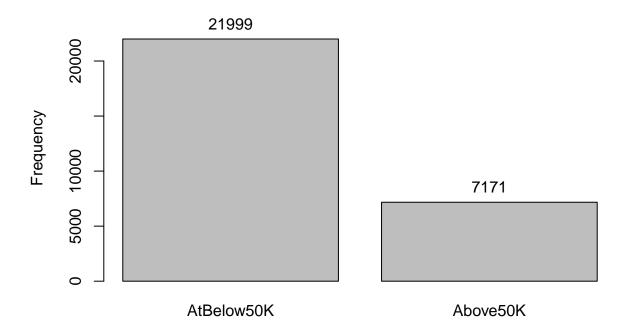


```
## adultpayclean$class :
                 Frequency Percent Cum. percent
##
## Private
                     20135
                               69.0
                                            69.0
                                7.9
                                            77.0
## Selfempnotinc
                      2313
## Localgov
                      1956
                                6.7
                                            83.7
## Unknown
                      1659
                                5.7
                                            89.3
## Stategov
                      1210
                                4.1
                                            93.5
## Selfempinc
                       991
                                3.4
                                            96.9
## Federalgov
                       886
                                3.0
                                            99.9
## Withoutpay
                        13
                                0.0
                                           100.0
## Neverworked
                         7
                                0.0
                                           100.0
##
     Total
                     29170
                              100.0
                                           100.0
adultpayclean %>% group_by(class) %>%
  mutate(n=n()) %>% ggplot() +
  geom_bar(aes(class,col=income,fill=income)) + scale_y_sqrt()
```



tab1(adultpayclean\$income, sort.group = "decreasing", cum.percent = TRUE)

# Distribution of adultpayclean\$income



```
## adultpayclean$income :

## Frequency Percent Cum. percent

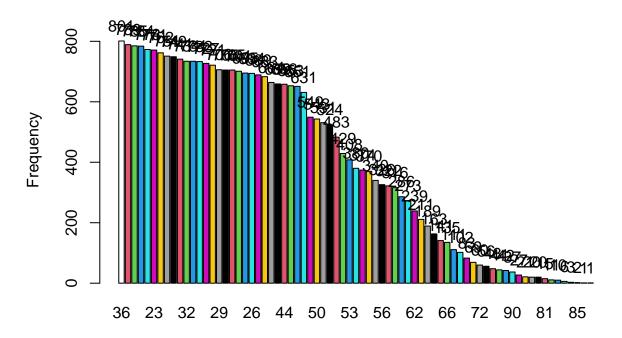
## AtBelow50K 21999 75.4 75.4

## Above50K 7171 24.6 100.0

## Total 29170 100.0 100.0
```

tab1(adultpayclean\$age, sort.group = "decreasing", cum.percent = TRUE)

# Distribution of adultpayclean\$age



##	adultpayclean\$age	:		
##	Frequency	Percent	Cum.	percent
##		2.7		2.7
##	35 789	2.7		5.5
##	31 785	2.7		8.1
##	34 784	2.7		10.8
##	33 773	2.6		13.5
##	23 771	2.6		16.1
##	37 762	2.6		18.7
##	28 751	2.6		21.3
##	30 749	2.6		23.9
##	38 741	2.5		26.4
##	32 734	2.5		28.9
##	25 734	2.5		31.5
##	39 733	2.5		34.0
##	27 727	2.5		36.5
##	40 721	2.5		38.9
##	29 706	2.4		41.3
##	41 705	2.4		43.8
##	24 705	2.4		46.2
##	42 701	2.4		48.6
##	43 695	2.4		51.0
##	26 694	2.4		53.3
##	20 689	2.4		55.7
##	22 683	2.3		58.0
##	46 664	2.3		60.3

```
## 19
                   659
                            2.3
                                          62.6
## 44
                   658
                            2.3
                                          64.8
## 45
                   653
                            2.2
                                          67.1
## 21
                   651
                            2.2
                                          69.3
## 47
                   631
                            2.2
                                          71.5
## 51
                   549
                            1.9
                                          73.4
## 50
                   543
                            1.9
                                          75.2
## 49
                   531
                                          77.0
                            1.8
## 18
                   524
                            1.8
                                          78.8
## 48
                            1.7
                                          80.5
                   483
## 52
                   429
                            1.5
                                          82.0
## 53
                   408
                            1.4
                                          83.4
## 55
                   380
                            1.3
                                          84.7
## 17
                   374
                                          85.9
                            1.3
## 54
                   370
                            1.3
                                          87.2
## 58
                   340
                            1.2
                                          88.4
## 56
                   326
                            1.1
                                          89.5
## 59
                   322
                            1.1
                                          90.6
## 57
                   316
                            1.1
                                          91.7
## 60
                   286
                            1.0
                                          92.7
## 61
                   273
                            0.9
                                          93.6
## 62
                   239
                            0.8
                                          94.4
## 63
                                          95.1
                   211
                            0.7
## 64
                   189
                            0.6
                                          95.8
## 65
                            0.6
                                          96.3
                   163
## 67
                   141
                            0.5
                                          96.8
## 66
                   135
                            0.5
                                          97.3
##
   68
                            0.4
                                          97.7
                   111
## 69
                                          98.0
                   102
                            0.3
## 70
                            0.3
                                          98.3
                    83
## 71
                    69
                            0.2
                                          98.5
## 72
                    60
                            0.2
                                          98.8
## 73
                            0.2
                    56
                                          98.9
## 74
                    48
                            0.2
                                          99.1
## 76
                    44
                            0.2
                                          99.3
## 75
                    42
                            0.1
                                          99.4
## 90
                    37
                            0.1
                                          99.5
## 77
                    27
                            0.1
                                          99.6
## 80
                    21
                            0.1
                                          99.7
## 79
                    20
                                          99.8
                            0.1
## 78
                    20
                            0.1
                                          99.8
## 81
                    15
                            0.1
                                          99.9
## 82
                    11
                            0.0
                                          99.9
## 84
                                         100.0
                    10
                            0.0
## 83
                     6
                            0.0
                                         100.0
## 88
                     3
                            0.0
                                         100.0
## 85
                     2
                                         100.0
                            0.0
## 87
                     1
                            0.0
                                         100.0
## 86
                     1
                            0.0
                                         100.0
                 29170
##
     Total
                          100.0
                                         100.0
```

```
adultpayclean %>% group_by(age) %>%
mutate(n=n()) %>% ggplot() +
geom_bar(aes(age,col=income,fill=income)) + scale_y_sqrt()
```

