

Ejercicios Día 3

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22/10/2018

1. Dataframes

*Busca los datasets “beaver1” y “beaver2” que contienen información sobre la temperatura corporal de dos castores. Añade una columna llamada “ID” al dataset beaver1 que tenga siempre el valor 1. De forma similar añade una columna “ID” al dataset beaver2 que tenga siempre el valor 2. A continuación concatena de forma vertical los dos dataframes y busca el subet de datos donde ambos Castores están activos.

```
ID = rep(1,nrow(beaver1))
beaver1 = cbind(beaver1,ID)
ID = rep(2,nrow(beaver2))
beaver2 = cbind(beaver2,ID)
beaver = rbind.data.frame(beaver1,beaver2)
beaver[beaver[, "activ"] == 1,]
```

```
##      day time  temp activ ID
## 54  346 1730 37.07      1  1
## 68  346 1950 37.10      1  1
## 80  346 2150 37.53      1  1
## 83  346 2230 37.25      1  1
## 86  346 2300 37.24      1  1
## 114 347   340 37.15      1  1
## 153 307 1550 37.98      1  2
## 154 307 1600 38.02      1  2
## 155 307 1610 38.00      1  2
## 156 307 1620 38.24      1  2
## 157 307 1630 38.10      1  2
## 158 307 1640 38.24      1  2
## 159 307 1650 38.11      1  2
## 160 307 1700 38.02      1  2
## 161 307 1710 38.11      1  2
## 162 307 1720 38.01      1  2
## 163 307 1730 37.91      1  2
## 164 307 1740 37.96      1  2
## 165 307 1750 38.03      1  2
## 166 307 1800 38.17      1  2
## 167 307 1810 38.19      1  2
## 168 307 1820 38.18      1  2
## 169 307 1830 38.15      1  2
## 170 307 1840 38.04      1  2
## 171 307 1850 37.96      1  2
## 172 307 1900 37.84      1  2
## 173 307 1910 37.83      1  2
## 174 307 1920 37.84      1  2
## 175 307 1930 37.74      1  2
## 176 307 1940 37.76      1  2
```

```
## 177 307 1950 37.76      1  2
## 178 307 2000 37.64      1  2
## 179 307 2010 37.63      1  2
## 180 307 2020 38.06      1  2
## 181 307 2030 38.19      1  2
## 182 307 2040 38.35      1  2
## 183 307 2050 38.25      1  2
## 184 307 2100 37.86      1  2
## 185 307 2110 37.95      1  2
## 186 307 2120 37.95      1  2
## 187 307 2130 37.76      1  2
## 188 307 2140 37.60      1  2
## 189 307 2150 37.89      1  2
## 190 307 2200 37.86      1  2
## 191 307 2210 37.71      1  2
## 192 307 2220 37.78      1  2
## 193 307 2230 37.82      1  2
## 194 307 2240 37.76      1  2
## 195 307 2250 37.81      1  2
## 196 307 2300 37.84      1  2
## 197 307 2310 38.01      1  2
## 198 307 2320 38.10      1  2
## 199 307 2330 38.15      1  2
## 200 307 2340 37.92      1  2
## 201 307 2350 37.64      1  2
## 202 308    0 37.70      1  2
## 203 308   10 37.46      1  2
## 204 308   20 37.41      1  2
## 205 308   30 37.46      1  2
## 206 308   40 37.56      1  2
## 207 308   50 37.55      1  2
## 208 308  100 37.75      1  2
## 209 308  110 37.76      1  2
## 210 308  120 37.73      1  2
## 211 308  130 37.77      1  2
## 212 308  140 38.01      1  2
## 213 308  150 38.04      1  2
## 214 308  200 38.07      1  2
```

* Vamos a trabajar con un ejemplo que viene por defecto en la instalación de R USArrests. Este data frame contiene la información para cada estado Americano de las tasas de criminales (por 100.000 habitantes). Los datos de las columnas se refieren a Asesinatos, violaciones y porcentaje de la población que vive en áreas urbanas. Los datos son de 1973. Contesta a las siguientes preguntas sobre los datos:

```
#Las dimensiones del dataframe
dim(USArrests)
```

```
## [1] 50  4
```

```
#La longitud del dataframe (filas o columnas)
length(USArrests)
```

```
## [1] 4
```

```
#Número de columnas
ncol(USArrests)
```

```
## [1] 4
```

```
#¿Cómo calcularías el número de filas?
nrow(USArrests)
```

```
## [1] 50
```

```
#Obtén el nombre de las filas y las columnas para este data frame
row.names.data.frame(USArrests)
```

```
## [1] "Alabama"      "Alaska"      "Arizona"     "Arkansas"
## [5] "California"   "Colorado"    "Connecticut" "Delaware"
## [9] "Florida"     "Georgia"     "Hawaii"      "Idaho"
## [13] "Illinois"    "Indiana"     "Iowa"        "Kansas"
## [17] "Kentucky"    "Louisiana"   "Maine"       "Maryland"
## [21] "Massachusetts" "Michigan"    "Minnesota"   "Mississippi"
## [25] "Missouri"    "Montana"     "Nebraska"    "Nevada"
## [29] "New Hampshire" "New Jersey"  "New Mexico"  "New York"
## [33] "North Carolina" "North Dakota" "Ohio"        "Oklahoma"
## [37] "Oregon"      "Pennsylvania" "Rhode Island" "South Carolina"
## [41] "South Dakota" "Tennessee"   "Texas"       "Utah"
## [45] "Vermont"     "Virginia"    "Washington"  "West Virginia"
## [49] "Wisconsin"   "Wyoming"
```

```
colnames(USArrests)
```

```
## [1] "Murder" "Assault" "UrbanPop" "Rape"
```

```
#Échale un vistazo a los datos, por ejemplo a las seis primeras filas
USArrests[1:6,]
```

```
##      Murder Assault UrbanPop Rape
## Alabama    13.2    236      58 21.2
## Alaska     10.0    263      48 44.5
## Arizona     8.1    294      80 31.0
## Arkansas    8.8    190      50 19.5
## California  9.0    276      91 40.6
## Colorado   7.9    204      78 38.7
```

```
#Ordena de forma decreciente las filas de nuestro data frame según el porcentaje de población
#en el área urbana. Para ello investiga la función order () y sus parámetros.
```

```
USArrests[order(USArrests$UrbanPop,decreasing = TRUE),]
```

```
##      Murder Assault UrbanPop Rape
## California    9.0    276      91 40.6
## New Jersey    7.4    159      89 18.8
## Rhode Island  3.4    174      87  8.3
## New York     11.1    254      86 26.1
## Massachusetts 4.4    149      85 16.3
## Hawaii        5.3     46      83 20.2
## Illinois     10.4    249      83 24.0
## Nevada       12.2    252      81 46.0
## Arizona       8.1    294      80 31.0
## Florida      15.4    335      80 31.9
```

## Texas	12.7	201	80	25.5
## Utah	3.2	120	80	22.9
## Colorado	7.9	204	78	38.7
## Connecticut	3.3	110	77	11.1
## Ohio	7.3	120	75	21.4
## Michigan	12.1	255	74	35.1
## Washington	4.0	145	73	26.2
## Delaware	5.9	238	72	15.8
## Pennsylvania	6.3	106	72	14.9
## Missouri	9.0	178	70	28.2
## New Mexico	11.4	285	70	32.1
## Oklahoma	6.6	151	68	20.0
## Maryland	11.3	300	67	27.8
## Oregon	4.9	159	67	29.3
## Kansas	6.0	115	66	18.0
## Louisiana	15.4	249	66	22.2
## Minnesota	2.7	72	66	14.9
## Wisconsin	2.6	53	66	10.8
## Indiana	7.2	113	65	21.0
## Virginia	8.5	156	63	20.7
## Nebraska	4.3	102	62	16.5
## Georgia	17.4	211	60	25.8
## Wyoming	6.8	161	60	15.6
## Tennessee	13.2	188	59	26.9
## Alabama	13.2	236	58	21.2
## Iowa	2.2	56	57	11.3
## New Hampshire	2.1	57	56	9.5
## Idaho	2.6	120	54	14.2
## Montana	6.0	109	53	16.4
## Kentucky	9.7	109	52	16.3
## Maine	2.1	83	51	7.8
## Arkansas	8.8	190	50	19.5
## Alaska	10.0	263	48	44.5
## South Carolina	14.4	279	48	22.5
## North Carolina	13.0	337	45	16.1
## South Dakota	3.8	86	45	12.8
## Mississippi	16.1	259	44	17.1
## North Dakota	0.8	45	44	7.3
## West Virginia	5.7	81	39	9.3
## Vermont	2.2	48	32	11.2

#¿Podrías añadir un segundo criterio de orden?, ¿cómo?

```
USArrests[order(USArrests$UrbanPop,USArrests$Rape,decreasing = TRUE),]
```

##	Murder	Assault	UrbanPop	Rape
## California	9.0	276	91	40.6
## New Jersey	7.4	159	89	18.8
## Rhode Island	3.4	174	87	8.3
## New York	11.1	254	86	26.1
## Massachusetts	4.4	149	85	16.3
## Illinois	10.4	249	83	24.0
## Hawaii	5.3	46	83	20.2
## Nevada	12.2	252	81	46.0
## Florida	15.4	335	80	31.9

```
## Arizona      8.1      294      80 31.0
## Texas        12.7     201      80 25.5
## Utah         3.2      120      80 22.9
## Colorado     7.9      204      78 38.7
## Connecticut  3.3      110      77 11.1
## Ohio         7.3      120      75 21.4
## Michigan     12.1     255      74 35.1
## Washington   4.0      145      73 26.2
## Delaware     5.9      238      72 15.8
## Pennsylvania 6.3      106      72 14.9
## New Mexico   11.4     285      70 32.1
## Missouri     9.0      178      70 28.2
## Oklahoma     6.6      151      68 20.0
## Oregon       4.9      159      67 29.3
## Maryland     11.3     300      67 27.8
## Louisiana    15.4     249      66 22.2
## Kansas       6.0      115      66 18.0
## Minnesota    2.7       72      66 14.9
## Wisconsin    2.6       53      66 10.8
## Indiana      7.2      113      65 21.0
## Virginia     8.5      156      63 20.7
## Nebraska     4.3      102      62 16.5
## Georgia      17.4     211      60 25.8
## Wyoming      6.8      161      60 15.6
## Tennessee    13.2     188      59 26.9
## Alabama      13.2     236      58 21.2
## Iowa         2.2       56      57 11.3
## New Hampshire 2.1       57      56  9.5
## Idaho        2.6      120      54 14.2
## Montana      6.0      109      53 16.4
## Kentucky     9.7      109      52 16.3
## Maine        2.1       83      51  7.8
## Arkansas     8.8      190      50 19.5
## Alaska       10.0     263      48 44.5
## South Carolina 14.4     279      48 22.5
## North Carolina 13.0     337      45 16.1
## South Dakota  3.8       86      45 12.8
## Mississippi  16.1     259      44 17.1
## North Dakota  0.8       45      44  7.3
## West Virginia 5.7       81      39  9.3
## Vermont      2.2       48      32 11.2
```

```
#Muestra por pantalla la columna con los datos de asesinato
```

```
USArrests$Murder
```

```
## [1] 13.2 10.0  8.1  8.8  9.0  7.9  3.3  5.9 15.4 17.4  5.3  2.6 10.4  7.2
## [15]  2.2  6.0  9.7 15.4  2.1 11.3  4.4 12.1  2.7 16.1  9.0  6.0  4.3 12.2
## [29]  2.1  7.4 11.4 11.1 13.0  0.8  7.3  6.6  4.9  6.3  3.4 14.4  3.8 13.2
## [43] 12.7  3.2  2.2  8.5  4.0  5.7  2.6  6.8
```

```
#Muestra las tasas de asesinato para el segundo, tercer y cuarto estado
```

```
USArrests[2:4,"Murder"]
```

```
## [1] 10.0  8.1  8.8
```

#Muestra las primeras cinco filas de todas las columnas

```
USArrests[1:5,]
```

##		Murder	Assault	UrbanPop	Rape
##	Alabama	13.2	236	58	21.2
##	Alaska	10.0	263	48	44.5
##	Arizona	8.1	294	80	31.0
##	Arkansas	8.8	190	50	19.5
##	California	9.0	276	91	40.6

#Muestra todas las filas para las dos primeras columnas

```
USArrests[,1:2]
```

##		Murder	Assault
##	Alabama	13.2	236
##	Alaska	10.0	263
##	Arizona	8.1	294
##	Arkansas	8.8	190
##	California	9.0	276
##	Colorado	7.9	204
##	Connecticut	3.3	110
##	Delaware	5.9	238
##	Florida	15.4	335
##	Georgia	17.4	211
##	Hawaii	5.3	46
##	Idaho	2.6	120
##	Illinois	10.4	249
##	Indiana	7.2	113
##	Iowa	2.2	56
##	Kansas	6.0	115
##	Kentucky	9.7	109
##	Louisiana	15.4	249
##	Maine	2.1	83
##	Maryland	11.3	300
##	Massachusetts	4.4	149
##	Michigan	12.1	255
##	Minnesota	2.7	72
##	Mississippi	16.1	259
##	Missouri	9.0	178
##	Montana	6.0	109
##	Nebraska	4.3	102
##	Nevada	12.2	252
##	New Hampshire	2.1	57
##	New Jersey	7.4	159
##	New Mexico	11.4	285
##	New York	11.1	254
##	North Carolina	13.0	337
##	North Dakota	0.8	45
##	Ohio	7.3	120
##	Oklahoma	6.6	151
##	Oregon	4.9	159
##	Pennsylvania	6.3	106
##	Rhode Island	3.4	174

```
## South Carolina 14.4 279
## South Dakota 3.8 86
## Tennessee 13.2 188
## Texas 12.7 201
## Utah 3.2 120
## Vermont 2.2 48
## Virginia 8.5 156
## Washington 4.0 145
## West Virginia 5.7 81
## Wisconsin 2.6 53
## Wyoming 6.8 161
```

#Muestra todas las filas de las columnas 1 y 3

```
USArrests[,c(1,3)]
```

```
## Murder UrbanPop
## Alabama 13.2 58
## Alaska 10.0 48
## Arizona 8.1 80
## Arkansas 8.8 50
## California 9.0 91
## Colorado 7.9 78
## Connecticut 3.3 77
## Delaware 5.9 72
## Florida 15.4 80
## Georgia 17.4 60
## Hawaii 5.3 83
## Idaho 2.6 54
## Illinois 10.4 83
## Indiana 7.2 65
## Iowa 2.2 57
## Kansas 6.0 66
## Kentucky 9.7 52
## Louisiana 15.4 66
## Maine 2.1 51
## Maryland 11.3 67
## Massachusetts 4.4 85
## Michigan 12.1 74
## Minnesota 2.7 66
## Mississippi 16.1 44
## Missouri 9.0 70
## Montana 6.0 53
## Nebraska 4.3 62
## Nevada 12.2 81
## New Hampshire 2.1 56
## New Jersey 7.4 89
## New Mexico 11.4 70
## New York 11.1 86
## North Carolina 13.0 45
## North Dakota 0.8 44
## Ohio 7.3 75
## Oklahoma 6.6 68
## Oregon 4.9 67
## Pennsylvania 6.3 72
```

```
## Rhode Island      3.4      87
## South Carolina    14.4      48
## South Dakota       3.8      45
## Tennessee         13.2      59
## Texas             12.7      80
## Utah              3.2      80
## Vermont           2.2      32
## Virginia          8.5      63
## Washington        4.0      73
## West Virginia     5.7      39
## Wisconsin         2.6      66
## Wyoming           6.8      60
```

#Muestra solo las primeras cinco filas de las columnas 1 y 2

```
USArrests[1:5,1:2]
```

```
##           Murder Assault
## Alabama      13.2      236
## Alaska       10.0      263
## Arizona       8.1      294
## Arkansas      8.8      190
## California    9.0      276
```

#Extrae las filas para el índice Murder

```
USArrests[, "Murder"]
```

```
## [1] 13.2 10.0 8.1 8.8 9.0 7.9 3.3 5.9 15.4 17.4 5.3 2.6 10.4 7.2
## [15] 2.2 6.0 9.7 15.4 2.1 11.3 4.4 12.1 2.7 16.1 9.0 6.0 4.3 12.2
## [29] 2.1 7.4 11.4 11.1 13.0 0.8 7.3 6.6 4.9 6.3 3.4 14.4 3.8 13.2
## [43] 12.7 3.2 2.2 8.5 4.0 5.7 2.6 6.8
```

#¿Que estado tiene la menor tasa de asesinatos? ¿qué línea contiene esa información?, obtén esa información

```
datos = USArrests[order(USArrests$Murder,decreasing = FALSE),]
```

```
datos
```

```
##           Murder Assault UrbanPop Rape
## North Dakota    0.8      45      44  7.3
## Maine           2.1      83      51  7.8
## New Hampshire    2.1      57      56  9.5
## Iowa            2.2      56      57 11.3
## Vermont          2.2      48      32 11.2
## Idaho            2.6     120      54 14.2
## Wisconsin        2.6      53      66 10.8
## Minnesota        2.7      72      66 14.9
## Utah             3.2     120      80 22.9
## Connecticut      3.3     110      77 11.1
## Rhode Island     3.4     174      87  8.3
## South Dakota     3.8      86      45 12.8
## Washington       4.0     145      73 26.2
## Nebraska         4.3     102      62 16.5
## Massachusetts    4.4     149      85 16.3
## Oregon           4.9     159      67 29.3
## Hawaii           5.3      46      83 20.2
## West Virginia    5.7      81      39  9.3
## Delaware         5.9     238      72 15.8
## Kansas           6.0     115      66 18.0
```


## Montana	6.0	109	53	16.4
## Pennsylvania	6.3	106	72	14.9
## Oklahoma	6.6	151	68	20.0
## Wyoming	6.8	161	60	15.6
## Indiana	7.2	113	65	21.0
## Ohio	7.3	120	75	21.4
## New Jersey	7.4	159	89	18.8
## Colorado	7.9	204	78	38.7
## Arizona	8.1	294	80	31.0
## Virginia	8.5	156	63	20.7
## Arkansas	8.8	190	50	19.5
## California	9.0	276	91	40.6
## Missouri	9.0	178	70	28.2
## Kentucky	9.7	109	52	16.3
## Alaska	10.0	263	48	44.5
## Illinois	10.4	249	83	24.0
## New York	11.1	254	86	26.1
## Maryland	11.3	300	67	27.8
## New Mexico	11.4	285	70	32.1
## Michigan	12.1	255	74	35.1
## Nevada	12.2	252	81	46.0
## Texas	12.7	201	80	25.5
## North Carolina	13.0	337	45	16.1
## Alabama	13.2	236	58	21.2
## Tennessee	13.2	188	59	26.9
## South Carolina	14.4	279	48	22.5
## Florida	15.4	335	80	31.9
## Louisiana	15.4	249	66	22.2
## Mississippi	16.1	259	44	17.1
## Georgia	17.4	211	60	25.8

```
USArrests[which(USArrests == datos[1,"Murder"]),]
```

##	Murder	Assault	UrbanPop	Rape
## North Dakota	0.8	45	44	7.3

El estado del Norte de Dakota es el que menor tasa de asesinatos tiene y está en la posición 34 del dataset USArrests.

```
# ¿Qué estados tienen una tasa inferior al 4%?, obtén esa información
USArrests[which(USArrests$Murder < 4),]
```

##	Murder	Assault	UrbanPop	Rape
## Connecticut	3.3	110	77	11.1
## Idaho	2.6	120	54	14.2
## Iowa	2.2	56	57	11.3
## Maine	2.1	83	51	7.8
## Minnesota	2.7	72	66	14.9
## New Hampshire	2.1	57	56	9.5
## North Dakota	0.8	45	44	7.3
## Rhode Island	3.4	174	87	8.3
## South Dakota	3.8	86	45	12.8
## Utah	3.2	120	80	22.9
## Vermont	2.2	48	32	11.2
## Wisconsin	2.6	53	66	10.8

```
# Que estados estan en el cuartil superior (75) en lo que a poblacion en zonas urbanas se refiere
quantile(USArrests$UrbanPop)
```

```
##      0%    25%    50%    75%   100%
## 32.00 54.50 66.00 77.75 91.00
```

```
USArrests[which(USArrests$UrbanPop >= quantile(USArrests$UrbanPop)[4]) ,]
```

```
##           Murder Assault UrbanPop Rape
## Arizona         8.1      294      80 31.0
## California       9.0      276      91 40.6
## Colorado         7.9      204      78 38.7
## Florida          15.4     335      80 31.9
## Hawaii           5.3       46      83 20.2
## Illinois          10.4     249      83 24.0
## Massachusetts    4.4      149      85 16.3
## Nevada            12.2     252      81 46.0
## New Jersey        7.4      159      89 18.8
## New York          11.1     254      86 26.1
## Rhode Island      3.4      174      87  8.3
## Texas             12.7     201      80 25.5
## Utah              3.2      120      80 22.9
```

*Carga el set de datos CO2 y realiza las siguientes acciones:

a. Ordena alfabéticamente los datos en función de la variable Plant. Recuerda que Plant es un factor. Imprime el resultado por pantalla para comprobarlo.

```
rm(CO2)
```

```
## Warning in rm(CO2): object 'CO2' not found
```

```
factor(CO2$Plant)
```

```
## [1] Qn1 Qn1 Qn1 Qn1 Qn1 Qn1 Qn1 Qn1 Qn2 Qn2 Qn2 Qn2 Qn2 Qn2 Qn2 Qn3 Qn3 Qn3
## [18] Qn3 Qn3 Qn3 Qn3 Qc1 Qc1 Qc1 Qc1 Qc1 Qc1 Qc1 Qc1 Qc2 Qc2 Qc2 Qc2 Qc2 Qc2
## [35] Qc2 Qc3 Qc3 Qc3 Qc3 Qc3 Qc3 Qc3 Qc3 Mn1 Mn1 Mn1 Mn1 Mn1 Mn1 Mn1 Mn2 Mn2
## [52] Mn2 Mn2 Mn2 Mn2 Mn2 Mn3 Mn3 Mn3 Mn3 Mn3 Mn3 Mn3 Mn3 Mc1 Mc1 Mc1 Mc1 Mc1
## [69] Mc1 Mc1 Mc2 Mc2 Mc2 Mc2 Mc2 Mc2 Mc2 Mc2 Mc3 Mc3 Mc3 Mc3 Mc3 Mc3 Mc3
## 12 Levels: Qn1 < Qn2 < Qn3 < Qc1 < Qc3 < Qc2 < Mn3 < Mn2 < Mn1 < ... < Mc1
```

```
orden = levels(CO2$Plant)
```

```
c = orden[order(levels(CO2$Plant))]
```

```
levels(CO2$Plant) = c
```

```
CO2
```

```
##      Plant      Type Treatment conc uptake
## 1      Mc1      Quebec nonchilled    95    16.0
## 2      Mc1      Quebec nonchilled   175    30.4
## 3      Mc1      Quebec nonchilled   250    34.8
## 4      Mc1      Quebec nonchilled   350    37.2
## 5      Mc1      Quebec nonchilled   500    35.3
## 6      Mc1      Quebec nonchilled   675    39.2
## 7      Mc1      Quebec nonchilled  1000    39.7
## 8      Mc2      Quebec nonchilled    95    13.6
```

## 9	Mc2	Quebec nonchilled	175	27.3
## 10	Mc2	Quebec nonchilled	250	37.1
## 11	Mc2	Quebec nonchilled	350	41.8
## 12	Mc2	Quebec nonchilled	500	40.6
## 13	Mc2	Quebec nonchilled	675	41.4
## 14	Mc2	Quebec nonchilled	1000	44.3
## 15	Mc3	Quebec nonchilled	95	16.2
## 16	Mc3	Quebec nonchilled	175	32.4
## 17	Mc3	Quebec nonchilled	250	40.3
## 18	Mc3	Quebec nonchilled	350	42.1
## 19	Mc3	Quebec nonchilled	500	42.9
## 20	Mc3	Quebec nonchilled	675	43.9
## 21	Mc3	Quebec nonchilled	1000	45.5
## 22	Mn1	Quebec chilled	95	14.2
## 23	Mn1	Quebec chilled	175	24.1
## 24	Mn1	Quebec chilled	250	30.3
## 25	Mn1	Quebec chilled	350	34.6
## 26	Mn1	Quebec chilled	500	32.5
## 27	Mn1	Quebec chilled	675	35.4
## 28	Mn1	Quebec chilled	1000	38.7
## 29	Mn3	Quebec chilled	95	9.3
## 30	Mn3	Quebec chilled	175	27.3
## 31	Mn3	Quebec chilled	250	35.0
## 32	Mn3	Quebec chilled	350	38.8
## 33	Mn3	Quebec chilled	500	38.6
## 34	Mn3	Quebec chilled	675	37.5
## 35	Mn3	Quebec chilled	1000	42.4
## 36	Mn2	Quebec chilled	95	15.1
## 37	Mn2	Quebec chilled	175	21.0
## 38	Mn2	Quebec chilled	250	38.1
## 39	Mn2	Quebec chilled	350	34.0
## 40	Mn2	Quebec chilled	500	38.9
## 41	Mn2	Quebec chilled	675	39.6
## 42	Mn2	Quebec chilled	1000	41.4
## 43	Qc3	Mississippi nonchilled	95	10.6
## 44	Qc3	Mississippi nonchilled	175	19.2
## 45	Qc3	Mississippi nonchilled	250	26.2
## 46	Qc3	Mississippi nonchilled	350	30.0
## 47	Qc3	Mississippi nonchilled	500	30.9
## 48	Qc3	Mississippi nonchilled	675	32.4
## 49	Qc3	Mississippi nonchilled	1000	35.5
## 50	Qc2	Mississippi nonchilled	95	12.0
## 51	Qc2	Mississippi nonchilled	175	22.0
## 52	Qc2	Mississippi nonchilled	250	30.6
## 53	Qc2	Mississippi nonchilled	350	31.8
## 54	Qc2	Mississippi nonchilled	500	32.4
## 55	Qc2	Mississippi nonchilled	675	31.1
## 56	Qc2	Mississippi nonchilled	1000	31.5
## 57	Qc1	Mississippi nonchilled	95	11.3
## 58	Qc1	Mississippi nonchilled	175	19.4
## 59	Qc1	Mississippi nonchilled	250	25.8
## 60	Qc1	Mississippi nonchilled	350	27.9
## 61	Qc1	Mississippi nonchilled	500	28.5
## 62	Qc1	Mississippi nonchilled	675	28.1

```
## 63 Qc1 Mississippi nonchilled 1000 27.8
## 64 Qn3 Mississippi chilled 95 10.5
## 65 Qn3 Mississippi chilled 175 14.9
## 66 Qn3 Mississippi chilled 250 18.1
## 67 Qn3 Mississippi chilled 350 18.9
## 68 Qn3 Mississippi chilled 500 19.5
## 69 Qn3 Mississippi chilled 675 22.2
## 70 Qn3 Mississippi chilled 1000 21.9
## 71 Qn1 Mississippi chilled 95 7.7
## 72 Qn1 Mississippi chilled 175 11.4
## 73 Qn1 Mississippi chilled 250 12.3
## 74 Qn1 Mississippi chilled 350 13.0
## 75 Qn1 Mississippi chilled 500 12.5
## 76 Qn1 Mississippi chilled 675 13.7
## 77 Qn1 Mississippi chilled 1000 14.4
## 78 Qn2 Mississippi chilled 95 10.6
## 79 Qn2 Mississippi chilled 175 18.0
## 80 Qn2 Mississippi chilled 250 17.9
## 81 Qn2 Mississippi chilled 350 17.9
## 82 Qn2 Mississippi chilled 500 17.9
## 83 Qn2 Mississippi chilled 675 18.9
## 84 Qn2 Mississippi chilled 1000 19.9
```

b. Ordena los datos en función del incremento de la variable uptake y el orden alfabético de la planta (en ese orden)

```
C02[order(C02$uptake,C02$Plant),]
```

```
## Plant Type Treatment conc uptake
## 71 Qn1 Mississippi chilled 95 7.7
## 29 Mn3 Quebec chilled 95 9.3
## 64 Qn3 Mississippi chilled 95 10.5
## 43 Qc3 Mississippi nonchilled 95 10.6
## 78 Qn2 Mississippi chilled 95 10.6
## 57 Qc1 Mississippi nonchilled 95 11.3
## 72 Qn1 Mississippi chilled 175 11.4
## 50 Qc2 Mississippi nonchilled 95 12.0
## 73 Qn1 Mississippi chilled 250 12.3
## 75 Qn1 Mississippi chilled 500 12.5
## 74 Qn1 Mississippi chilled 350 13.0
## 8 Mc2 Quebec nonchilled 95 13.6
## 76 Qn1 Mississippi chilled 675 13.7
## 22 Mn1 Quebec chilled 95 14.2
## 77 Qn1 Mississippi chilled 1000 14.4
## 65 Qn3 Mississippi chilled 175 14.9
## 36 Mn2 Quebec chilled 95 15.1
## 1 Mc1 Quebec nonchilled 95 16.0
## 15 Mc3 Quebec nonchilled 95 16.2
## 80 Qn2 Mississippi chilled 250 17.9
## 81 Qn2 Mississippi chilled 350 17.9
## 82 Qn2 Mississippi chilled 500 17.9
## 79 Qn2 Mississippi chilled 175 18.0
## 66 Qn3 Mississippi chilled 250 18.1
```

## 83	Qn2	Mississippi	chilled	675	18.9
## 67	Qn3	Mississippi	chilled	350	18.9
## 44	Qc3	Mississippi	nonchilled	175	19.2
## 58	Qc1	Mississippi	nonchilled	175	19.4
## 68	Qn3	Mississippi	chilled	500	19.5
## 84	Qn2	Mississippi	chilled	1000	19.9
## 37	Mn2	Quebec	chilled	175	21.0
## 70	Qn3	Mississippi	chilled	1000	21.9
## 51	Qc2	Mississippi	nonchilled	175	22.0
## 69	Qn3	Mississippi	chilled	675	22.2
## 23	Mn1	Quebec	chilled	175	24.1
## 59	Qc1	Mississippi	nonchilled	250	25.8
## 45	Qc3	Mississippi	nonchilled	250	26.2
## 9	Mc2	Quebec	nonchilled	175	27.3
## 30	Mn3	Quebec	chilled	175	27.3
## 63	Qc1	Mississippi	nonchilled	1000	27.8
## 60	Qc1	Mississippi	nonchilled	350	27.9
## 62	Qc1	Mississippi	nonchilled	675	28.1
## 61	Qc1	Mississippi	nonchilled	500	28.5
## 46	Qc3	Mississippi	nonchilled	350	30.0
## 24	Mn1	Quebec	chilled	250	30.3
## 2	Mc1	Quebec	nonchilled	175	30.4
## 52	Qc2	Mississippi	nonchilled	250	30.6
## 47	Qc3	Mississippi	nonchilled	500	30.9
## 55	Qc2	Mississippi	nonchilled	675	31.1
## 56	Qc2	Mississippi	nonchilled	1000	31.5
## 53	Qc2	Mississippi	nonchilled	350	31.8
## 16	Mc3	Quebec	nonchilled	175	32.4
## 54	Qc2	Mississippi	nonchilled	500	32.4
## 48	Qc3	Mississippi	nonchilled	675	32.4
## 26	Mn1	Quebec	chilled	500	32.5
## 39	Mn2	Quebec	chilled	350	34.0
## 25	Mn1	Quebec	chilled	350	34.6
## 3	Mc1	Quebec	nonchilled	250	34.8
## 31	Mn3	Quebec	chilled	250	35.0
## 5	Mc1	Quebec	nonchilled	500	35.3
## 27	Mn1	Quebec	chilled	675	35.4
## 49	Qc3	Mississippi	nonchilled	1000	35.5
## 10	Mc2	Quebec	nonchilled	250	37.1
## 4	Mc1	Quebec	nonchilled	350	37.2
## 34	Mn3	Quebec	chilled	675	37.5
## 38	Mn2	Quebec	chilled	250	38.1
## 33	Mn3	Quebec	chilled	500	38.6
## 28	Mn1	Quebec	chilled	1000	38.7
## 32	Mn3	Quebec	chilled	350	38.8
## 40	Mn2	Quebec	chilled	500	38.9
## 6	Mc1	Quebec	nonchilled	675	39.2
## 41	Mn2	Quebec	chilled	675	39.6
## 7	Mc1	Quebec	nonchilled	1000	39.7
## 17	Mc3	Quebec	nonchilled	250	40.3
## 12	Mc2	Quebec	nonchilled	500	40.6
## 13	Mc2	Quebec	nonchilled	675	41.4
## 42	Mn2	Quebec	chilled	1000	41.4
## 11	Mc2	Quebec	nonchilled	350	41.8

```
## 18  Mc3      Quebec nonchilled 350 42.1
## 35  Mn3      Quebec   chilled 1000 42.4
## 19  Mc3      Quebec nonchilled 500 42.9
## 20  Mc3      Quebec nonchilled 675 43.9
## 14  Mc2      Quebec nonchilled 1000 44.3
## 21  Mc3      Quebec nonchilled 1000 45.5
```

Para observar que se ha ordenado de forma correcta, bastaría con buscar dos valores idénticos de uptake y comprobar si está ordenado de forma correcta por el tipo de planta estando los índices descolocados.

c. Ordena de nuevo los datos en function del increment de la variable uptake y el orden alfabético reverso de la planta (en ese orden)

```
C02[order(C02$uptake,order(C02$Plant, decreasing = TRUE)),]
```

```
##      Plant      Type Treatment conc uptake
## 71  Qn1 Mississippi chilled    95    7.7
## 29  Mn3      Quebec chilled    95    9.3
## 64  Qn3 Mississippi chilled    95   10.5
## 78  Qn2 Mississippi chilled    95   10.6
## 43  Qc3 Mississippi nonchilled  95   10.6
## 57  Qc1 Mississippi nonchilled  95   11.3
## 72  Qn1 Mississippi chilled   175   11.4
## 50  Qc2 Mississippi nonchilled  95   12.0
## 73  Qn1 Mississippi chilled   250   12.3
## 75  Qn1 Mississippi chilled   500   12.5
## 74  Qn1 Mississippi chilled   350   13.0
## 8   Mc2      Quebec nonchilled  95   13.6
## 76  Qn1 Mississippi chilled   675   13.7
## 22  Mn1      Quebec chilled    95   14.2
## 77  Qn1 Mississippi chilled  1000   14.4
## 65  Qn3 Mississippi chilled   175   14.9
## 36  Mn2      Quebec chilled    95   15.1
## 1   Mc1      Quebec nonchilled  95   16.0
## 15  Mc3      Quebec nonchilled  95   16.2
## 80  Qn2 Mississippi chilled   250   17.9
## 81  Qn2 Mississippi chilled   350   17.9
## 82  Qn2 Mississippi chilled   500   17.9
## 79  Qn2 Mississippi chilled   175   18.0
## 66  Qn3 Mississippi chilled   250   18.1
## 83  Qn2 Mississippi chilled   675   18.9
## 67  Qn3 Mississippi chilled   350   18.9
## 44  Qc3 Mississippi nonchilled  175   19.2
## 58  Qc1 Mississippi nonchilled  175   19.4
## 68  Qn3 Mississippi chilled   500   19.5
## 84  Qn2 Mississippi chilled  1000   19.9
## 37  Mn2      Quebec chilled   175   21.0
## 70  Qn3 Mississippi chilled  1000   21.9
## 51  Qc2 Mississippi nonchilled  175   22.0
## 69  Qn3 Mississippi chilled   675   22.2
## 23  Mn1      Quebec chilled   175   24.1
## 59  Qc1 Mississippi nonchilled  250   25.8
## 45  Qc3 Mississippi nonchilled  250   26.2
```

##	30	Mn3	Quebec	chilled	175	27.3
##	9	Mc2	Quebec	nonchilled	175	27.3
##	63	Qc1	Mississippi	nonchilled	1000	27.8
##	60	Qc1	Mississippi	nonchilled	350	27.9
##	62	Qc1	Mississippi	nonchilled	675	28.1
##	61	Qc1	Mississippi	nonchilled	500	28.5
##	46	Qc3	Mississippi	nonchilled	350	30.0
##	24	Mn1	Quebec	chilled	250	30.3
##	2	Mc1	Quebec	nonchilled	175	30.4
##	52	Qc2	Mississippi	nonchilled	250	30.6
##	47	Qc3	Mississippi	nonchilled	500	30.9
##	55	Qc2	Mississippi	nonchilled	675	31.1
##	56	Qc2	Mississippi	nonchilled	1000	31.5
##	53	Qc2	Mississippi	nonchilled	350	31.8
##	48	Qc3	Mississippi	nonchilled	675	32.4
##	54	Qc2	Mississippi	nonchilled	500	32.4
##	16	Mc3	Quebec	nonchilled	175	32.4
##	26	Mn1	Quebec	chilled	500	32.5
##	39	Mn2	Quebec	chilled	350	34.0
##	25	Mn1	Quebec	chilled	350	34.6
##	3	Mc1	Quebec	nonchilled	250	34.8
##	31	Mn3	Quebec	chilled	250	35.0
##	5	Mc1	Quebec	nonchilled	500	35.3
##	27	Mn1	Quebec	chilled	675	35.4
##	49	Qc3	Mississippi	nonchilled	1000	35.5
##	10	Mc2	Quebec	nonchilled	250	37.1
##	4	Mc1	Quebec	nonchilled	350	37.2
##	34	Mn3	Quebec	chilled	675	37.5
##	38	Mn2	Quebec	chilled	250	38.1
##	33	Mn3	Quebec	chilled	500	38.6
##	28	Mn1	Quebec	chilled	1000	38.7
##	32	Mn3	Quebec	chilled	350	38.8
##	40	Mn2	Quebec	chilled	500	38.9
##	6	Mc1	Quebec	nonchilled	675	39.2
##	41	Mn2	Quebec	chilled	675	39.6
##	7	Mc1	Quebec	nonchilled	1000	39.7
##	17	Mc3	Quebec	nonchilled	250	40.3
##	12	Mc2	Quebec	nonchilled	500	40.6
##	42	Mn2	Quebec	chilled	1000	41.4
##	13	Mc2	Quebec	nonchilled	675	41.4
##	11	Mc2	Quebec	nonchilled	350	41.8
##	18	Mc3	Quebec	nonchilled	350	42.1
##	35	Mn3	Quebec	chilled	1000	42.4
##	19	Mc3	Quebec	nonchilled	500	42.9
##	20	Mc3	Quebec	nonchilled	675	43.9
##	14	Mc2	Quebec	nonchilled	1000	44.3
##	21	Mc3	Quebec	nonchilled	1000	45.5

* Para este ejercicio vamos a usar el dataset state.x77. Asegurate de que el objeto es un dataframe, si no lo es fuerza su conversión.

```
class(state.x77)
```

```
## [1] "matrix"
```

```
state.x77.data.frame = as.data.frame(state.x77)
class(state.x77.data.frame)
```

```
## [1] "data.frame"
```

-Averigua cuantos estados tienen ingresos (Income) menores de 4300. Pista investiga subset()

```
subset(state.x77.data.frame, Income < 4300)
```

```
##           Population Income Illiteracy Life Exp Murder HS Grad Frost
## Alabama           3615   3624         2.1   69.05   15.1   41.3    20
## Arkansas          2110   3378         1.9   70.66   10.1   39.9    65
## Georgia           4931   4091         2.0   68.54   13.9   40.6    60
## Idaho              813   4119         0.6   71.87    5.3   59.5   126
## Kentucky          3387   3712         1.6   70.10   10.6   38.5    95
## Louisiana          3806   3545         2.8   68.76   13.2   42.2    12
## Maine             1058   3694         0.7   70.39    2.7   54.7   161
## Mississippi        2341   3098         2.4   68.09   12.5   41.0    50
## Missouri           4767   4254         0.8   70.69    9.3   48.8   108
## New Hampshire       812   4281         0.7   71.23    3.3   57.6   174
## New Mexico         1144   3601         2.2   70.32    9.7   55.2   120
## North Carolina     5441   3875         1.8   69.21   11.1   38.5    80
## Oklahoma            2715   3983         1.1   71.42    6.4   51.6    82
## South Carolina      2816   3635         2.3   67.96   11.6   37.8    65
## South Dakota         681   4167         0.5   72.08    1.7   53.3   172
## Tennessee           4173   3821         1.7   70.11   11.0   41.8    70
## Texas             12237   4188         2.2   70.90   12.2   47.4    35
## Utah               1203   4022         0.6   72.90    4.5   67.3   137
## Vermont             472   3907         0.6   71.64    5.5   57.1   168
## West Virginia      1799   3617         1.4   69.48    6.7   41.6   100
##
##           Area
## Alabama      50708
## Arkansas     51945
## Georgia      58073
## Idaho        82677
## Kentucky     39650
## Louisiana    44930
## Maine        30920
## Mississippi  47296
## Missouri     68995
## New Hampshire 9027
## New Mexico   121412
## North Carolina 48798
## Oklahoma     68782
## South Carolina 30225
## South Dakota  75955
## Tennessee    41328
## Texas        262134
## Utah         82096
## Vermont       9267
## West Virginia 24070
```


-Averigua cual es el estado con los ingresos mas altos.

```
max(state.x77.data.frame$Income)

## [1] 6315

state.x77.data.frame[state.x77.data.frame$Income==max(state.x77.data.frame$Income),]

##      Population Income Illiteracy Life Exp Murder HS Grad Frost   Area
## Alaska      365    6315         1.5   69.31   11.3   66.7   152 566432
```

-Crea un data frame 2 df2 con los datasets existentes en R: state.abb, state.area, state.division, state.name, state.region. Las filas tienen que ser los nombres de los estados.

```
df2 = data.frame(state.abb,state.area,state.division, state.region, row.names = state.name)
df2
```

	state.abb	state.area	state.division	state.region
## Alabama	AL	51609	East South Central	South
## Alaska	AK	589757	Pacific	West
## Arizona	AZ	113909	Mountain	West
## Arkansas	AR	53104	West South Central	South
## California	CA	158693	Pacific	West
## Colorado	CO	104247	Mountain	West
## Connecticut	CT	5009	New England	Northeast
## Delaware	DE	2057	South Atlantic	South
## Florida	FL	58560	South Atlantic	South
## Georgia	GA	58876	South Atlantic	South
## Hawaii	HI	6450	Pacific	West
## Idaho	ID	83557	Mountain	West
## Illinois	IL	56400	East North Central	North Central
## Indiana	IN	36291	East North Central	North Central
## Iowa	IA	56290	West North Central	North Central
## Kansas	KS	82264	West North Central	North Central
## Kentucky	KY	40395	East South Central	South
## Louisiana	LA	48523	West South Central	South
## Maine	ME	33215	New England	Northeast
## Maryland	MD	10577	South Atlantic	South
## Massachusetts	MA	8257	New England	Northeast
## Michigan	MI	58216	East North Central	North Central
## Minnesota	MN	84068	West North Central	North Central
## Mississippi	MS	47716	East South Central	South
## Missouri	MO	69686	West North Central	North Central
## Montana	MT	147138	Mountain	West
## Nebraska	NE	77227	West North Central	North Central
## Nevada	NV	110540	Mountain	West
## New Hampshire	NH	9304	New England	Northeast
## New Jersey	NJ	7836	Middle Atlantic	Northeast
## New Mexico	NM	121666	Mountain	West
## New York	NY	49576	Middle Atlantic	Northeast
## North Carolina	NC	52586	South Atlantic	South
## North Dakota	ND	70665	West North Central	North Central
## Ohio	OH	41222	East North Central	North Central
## Oklahoma	OK	69919	West South Central	South

## Oregon	OR	96981	Pacific	West
## Pennsylvania	PA	45333	Middle Atlantic	Northeast
## Rhode Island	RI	1214	New England	Northeast
## South Carolina	SC	31055	South Atlantic	South
## South Dakota	SD	77047	West North Central	North Central
## Tennessee	TN	42244	East South Central	South
## Texas	TX	267339	West South Central	South
## Utah	UT	84916	Mountain	West
## Vermont	VT	9609	New England	Northeast
## Virginia	VA	40815	South Atlantic	South
## Washington	WA	68192	Pacific	West
## West Virginia	WV	24181	South Atlantic	South
## Wisconsin	WI	56154	East North Central	North Central
## Wyoming	WY	97914	Mountain	West

- Elimina de todas las variables la palabra state. Busca alguna función para strings.

```
colnames(df2) =unlist(strsplit(colnames(df2), split="state."))[-seq(1,8,by= 2)]
df2
```

##	abb	area	division	region
## Alabama	AL	51609	East South Central	South
## Alaska	AK	589757	Pacific	West
## Arizona	AZ	113909	Mountain	West
## Arkansas	AR	53104	West South Central	South
## California	CA	158693	Pacific	West
## Colorado	CO	104247	Mountain	West
## Connecticut	CT	5009	New England	Northeast
## Delaware	DE	2057	South Atlantic	South
## Florida	FL	58560	South Atlantic	South
## Georgia	GA	58876	South Atlantic	South
## Hawaii	HI	6450	Pacific	West
## Idaho	ID	83557	Mountain	West
## Illinois	IL	56400	East North Central	North Central
## Indiana	IN	36291	East North Central	North Central
## Iowa	IA	56290	West North Central	North Central
## Kansas	KS	82264	West North Central	North Central
## Kentucky	KY	40395	East South Central	South
## Louisiana	LA	48523	West South Central	South
## Maine	ME	33215	New England	Northeast
## Maryland	MD	10577	South Atlantic	South
## Massachusetts	MA	8257	New England	Northeast
## Michigan	MI	58216	East North Central	North Central
## Minnesota	MN	84068	West North Central	North Central
## Mississippi	MS	47716	East South Central	South
## Missouri	MO	69686	West North Central	North Central
## Montana	MT	147138	Mountain	West
## Nebraska	NE	77227	West North Central	North Central
## Nevada	NV	110540	Mountain	West
## New Hampshire	NH	9304	New England	Northeast
## New Jersey	NJ	7836	Middle Atlantic	Northeast
## New Mexico	NM	121666	Mountain	West
## New York	NY	49576	Middle Atlantic	Northeast

## North Carolina	NC	52586	South Atlantic	South
## North Dakota	ND	70665	West North Central	North Central
## Ohio	OH	41222	East North Central	North Central
## Oklahoma	OK	69919	West South Central	South
## Oregon	OR	96981	Pacific	West
## Pennsylvania	PA	45333	Middle Atlantic	Northeast
## Rhode Island	RI	1214	New England	Northeast
## South Carolina	SC	31055	South Atlantic	South
## South Dakota	SD	77047	West North Central	North Central
## Tennessee	TN	42244	East South Central	South
## Texas	TX	267339	West South Central	South
## Utah	UT	84916	Mountain	West
## Vermont	VT	9609	New England	Northeast
## Virginia	VA	40815	South Atlantic	South
## Washington	WA	68192	Pacific	West
## West Virginia	WV	24181	South Atlantic	South
## Wisconsin	WI	56154	East North Central	North Central
## Wyoming	WY	97914	Mountain	West

-Borra la variable div de df2. Estas borrando una ??nica variable del dataframe

```
df2 = df2[,-3]
```

-Añade por columnas el nuevo dataframe df2 al dataframe state.x77. Elimina las variables Life Exp, HS Grad, Frost, abb, y are.

```
state.x77.data.frame = cbind.data.frame(state.x77,df2)
state.x77.data.frame
```

##	Population	Income	Illiteracy	Life Exp	Murder	HS Grad	Frost
## Alabama	3615	3624	2.1	69.05	15.1	41.3	20
## Alaska	365	6315	1.5	69.31	11.3	66.7	152
## Arizona	2212	4530	1.8	70.55	7.8	58.1	15
## Arkansas	2110	3378	1.9	70.66	10.1	39.9	65
## California	21198	5114	1.1	71.71	10.3	62.6	20
## Colorado	2541	4884	0.7	72.06	6.8	63.9	166
## Connecticut	3100	5348	1.1	72.48	3.1	56.0	139
## Delaware	579	4809	0.9	70.06	6.2	54.6	103
## Florida	8277	4815	1.3	70.66	10.7	52.6	11
## Georgia	4931	4091	2.0	68.54	13.9	40.6	60
## Hawaii	868	4963	1.9	73.60	6.2	61.9	0
## Idaho	813	4119	0.6	71.87	5.3	59.5	126
## Illinois	11197	5107	0.9	70.14	10.3	52.6	127
## Indiana	5313	4458	0.7	70.88	7.1	52.9	122
## Iowa	2861	4628	0.5	72.56	2.3	59.0	140
## Kansas	2280	4669	0.6	72.58	4.5	59.9	114
## Kentucky	3387	3712	1.6	70.10	10.6	38.5	95
## Louisiana	3806	3545	2.8	68.76	13.2	42.2	12
## Maine	1058	3694	0.7	70.39	2.7	54.7	161
## Maryland	4122	5299	0.9	70.22	8.5	52.3	101
## Massachusetts	5814	4755	1.1	71.83	3.3	58.5	103
## Michigan	9111	4751	0.9	70.63	11.1	52.8	125

## Minnesota	3921	4675	0.6	72.96	2.3	57.6	160
## Mississippi	2341	3098	2.4	68.09	12.5	41.0	50
## Missouri	4767	4254	0.8	70.69	9.3	48.8	108
## Montana	746	4347	0.6	70.56	5.0	59.2	155
## Nebraska	1544	4508	0.6	72.60	2.9	59.3	139
## Nevada	590	5149	0.5	69.03	11.5	65.2	188
## New Hampshire	812	4281	0.7	71.23	3.3	57.6	174
## New Jersey	7333	5237	1.1	70.93	5.2	52.5	115
## New Mexico	1144	3601	2.2	70.32	9.7	55.2	120
## New York	18076	4903	1.4	70.55	10.9	52.7	82
## North Carolina	5441	3875	1.8	69.21	11.1	38.5	80
## North Dakota	637	5087	0.8	72.78	1.4	50.3	186
## Ohio	10735	4561	0.8	70.82	7.4	53.2	124
## Oklahoma	2715	3983	1.1	71.42	6.4	51.6	82
## Oregon	2284	4660	0.6	72.13	4.2	60.0	44
## Pennsylvania	11860	4449	1.0	70.43	6.1	50.2	126
## Rhode Island	931	4558	1.3	71.90	2.4	46.4	127
## South Carolina	2816	3635	2.3	67.96	11.6	37.8	65
## South Dakota	681	4167	0.5	72.08	1.7	53.3	172
## Tennessee	4173	3821	1.7	70.11	11.0	41.8	70
## Texas	12237	4188	2.2	70.90	12.2	47.4	35
## Utah	1203	4022	0.6	72.90	4.5	67.3	137
## Vermont	472	3907	0.6	71.64	5.5	57.1	168
## Virginia	4981	4701	1.4	70.08	9.5	47.8	85
## Washington	3559	4864	0.6	71.72	4.3	63.5	32
## West Virginia	1799	3617	1.4	69.48	6.7	41.6	100
## Wisconsin	4589	4468	0.7	72.48	3.0	54.5	149
## Wyoming	376	4566	0.6	70.29	6.9	62.9	173
##	Area	abb	area	region			
## Alabama	50708	AL	51609	South			
## Alaska	566432	AK	589757	West			
## Arizona	113417	AZ	113909	West			
## Arkansas	51945	AR	53104	South			
## California	156361	CA	158693	West			
## Colorado	103766	CO	104247	West			
## Connecticut	4862	CT	5009	Northeast			
## Delaware	1982	DE	2057	South			
## Florida	54090	FL	58560	South			
## Georgia	58073	GA	58876	South			
## Hawaii	6425	HI	6450	West			
## Idaho	82677	ID	83557	West			
## Illinois	55748	IL	56400	North Central			
## Indiana	36097	IN	36291	North Central			
## Iowa	55941	IA	56290	North Central			
## Kansas	81787	KS	82264	North Central			
## Kentucky	39650	KY	40395	South			
## Louisiana	44930	LA	48523	South			
## Maine	30920	ME	33215	Northeast			
## Maryland	9891	MD	10577	South			
## Massachusetts	7826	MA	8257	Northeast			
## Michigan	56817	MI	58216	North Central			
## Minnesota	79289	MN	84068	North Central			
## Mississippi	47296	MS	47716	South			
## Missouri	68995	MO	69686	North Central			

```
## Montana      145587 MT 147138      West
## Nebraska      76483 NE 77227 North Central
## Nevada      109889 NV 110540      West
## New Hampshire  9027  NH  9304      Northeast
## New Jersey     7521  NJ  7836      Northeast
## New Mexico    121412 NM 121666      West
## New York      47831 NY 49576      Northeast
## North Carolina 48798 NC 52586      South
## North Dakota   69273 ND 70665 North Central
## Ohio          40975 OH 41222 North Central
## Oklahoma      68782 OK 69919      South
## Oregon        96184 OR 96981      West
## Pennsylvania  44966 PA 45333      Northeast
## Rhode Island   1049  RI  1214      Northeast
## South Carolina 30225 SC 31055      South
## South Dakota   75955 SD 77047 North Central
## Tennessee     41328 TN 42244      South
## Texas         262134 TX 267339      South
## Utah          82096 UT 84916      West
## Vermont       9267  VT  9609      Northeast
## Virginia      39780 VA 40815      South
## Washington    66570 WA 68192      West
## West Virginia  24070 WV 24181      South
## Wisconsin     54464 WI 56154 North Central
## Wyoming       97203 WY 97914      West
```

```
state.x77.data.frame =state.x77.data.frame[, -c(4,6,7,9,10)]
state.x77.data.frame
```

```
##      Population Income Illiteracy Murder   Area      region
## Alabama      3615   3624      2.1   15.1  50708      South
## Alaska       365   6315      1.5   11.3 566432      West
## Arizona      2212   4530      1.8    7.8 113417      West
## Arkansas     2110   3378      1.9   10.1  51945      South
## California   21198   5114      1.1   10.3 156361      West
## Colorado     2541   4884      0.7    6.8 103766      West
## Connecticut  3100   5348      1.1    3.1   4862 Northeast
## Delaware      579   4809      0.9    6.2   1982      South
## Florida      8277   4815      1.3   10.7  54090      South
## Georgia      4931   4091      2.0   13.9  58073      South
## Hawaii       868   4963      1.9    6.2   6425      West
## Idaho        813   4119      0.6    5.3  82677      West
## Illinois     11197   5107      0.9   10.3  55748 North Central
## Indiana      5313   4458      0.7    7.1  36097 North Central
## Iowa        2861   4628      0.5    2.3  55941 North Central
## Kansas       2280   4669      0.6    4.5  81787 North Central
## Kentucky     3387   3712      1.6   10.6  39650      South
## Louisiana    3806   3545      2.8   13.2  44930      South
## Maine        1058   3694      0.7    2.7  30920 Northeast
## Maryland     4122   5299      0.9    8.5   9891      South
## Massachusetts 5814   4755      1.1    3.3   7826 Northeast
## Michigan     9111   4751      0.9   11.1  56817 North Central
## Minnesota    3921   4675      0.6    2.3  79289 North Central
## Mississippi  2341   3098      2.4   12.5  47296      South
## Missouri     4767   4254      0.8    9.3  68995 North Central
```

## Montana	746	4347	0.6	5.0	145587	West
## Nebraska	1544	4508	0.6	2.9	76483	North Central
## Nevada	590	5149	0.5	11.5	109889	West
## New Hampshire	812	4281	0.7	3.3	9027	Northeast
## New Jersey	7333	5237	1.1	5.2	7521	Northeast
## New Mexico	1144	3601	2.2	9.7	121412	West
## New York	18076	4903	1.4	10.9	47831	Northeast
## North Carolina	5441	3875	1.8	11.1	48798	South
## North Dakota	637	5087	0.8	1.4	69273	North Central
## Ohio	10735	4561	0.8	7.4	40975	North Central
## Oklahoma	2715	3983	1.1	6.4	68782	South
## Oregon	2284	4660	0.6	4.2	96184	West
## Pennsylvania	11860	4449	1.0	6.1	44966	Northeast
## Rhode Island	931	4558	1.3	2.4	1049	Northeast
## South Carolina	2816	3635	2.3	11.6	30225	South
## South Dakota	681	4167	0.5	1.7	75955	North Central
## Tennessee	4173	3821	1.7	11.0	41328	South
## Texas	12237	4188	2.2	12.2	262134	South
## Utah	1203	4022	0.6	4.5	82096	West
## Vermont	472	3907	0.6	5.5	9267	Northeast
## Virginia	4981	4701	1.4	9.5	39780	South
## Washington	3559	4864	0.6	4.3	66570	West
## West Virginia	1799	3617	1.4	6.7	24070	South
## Wisconsin	4589	4468	0.7	3.0	54464	North Central
## Wyoming	376	4566	0.6	6.9	97203	West

-Añade una variable que categorice el nivel de formacion (illiteracy) de manera que [0,1) is low, [1,2) is some, [2, inf) is high.Pista. Hazlo de dos formas usando la función cut() y usando ifelse().

```
#Usando cut
categorizar.cut = cut(state.x77.data.frame$Illiteracy, breaks =c(0,1,2,Inf), labels=c("low", "some", "high"))

state.x77.data.frame = data.frame(state.x77.data.frame,categorizar.cut)
state.x77.data.frame
```

##	Population	Income	Illiteracy	Murder	Area	region
## Alabama	3615	3624	2.1	15.1	50708	South
## Alaska	365	6315	1.5	11.3	566432	West
## Arizona	2212	4530	1.8	7.8	113417	West
## Arkansas	2110	3378	1.9	10.1	51945	South
## California	21198	5114	1.1	10.3	156361	West
## Colorado	2541	4884	0.7	6.8	103766	West
## Connecticut	3100	5348	1.1	3.1	4862	Northeast
## Delaware	579	4809	0.9	6.2	1982	South
## Florida	8277	4815	1.3	10.7	54090	South
## Georgia	4931	4091	2.0	13.9	58073	South
## Hawaii	868	4963	1.9	6.2	6425	West
## Idaho	813	4119	0.6	5.3	82677	West
## Illinois	11197	5107	0.9	10.3	55748	North Central
## Indiana	5313	4458	0.7	7.1	36097	North Central
## Iowa	2861	4628	0.5	2.3	55941	North Central
## Kansas	2280	4669	0.6	4.5	81787	North Central

## Kentucky	3387	3712	1.6	10.6	39650	South
## Louisiana	3806	3545	2.8	13.2	44930	South
## Maine	1058	3694	0.7	2.7	30920	Northeast
## Maryland	4122	5299	0.9	8.5	9891	South
## Massachusetts	5814	4755	1.1	3.3	7826	Northeast
## Michigan	9111	4751	0.9	11.1	56817	North Central
## Minnesota	3921	4675	0.6	2.3	79289	North Central
## Mississippi	2341	3098	2.4	12.5	47296	South
## Missouri	4767	4254	0.8	9.3	68995	North Central
## Montana	746	4347	0.6	5.0	145587	West
## Nebraska	1544	4508	0.6	2.9	76483	North Central
## Nevada	590	5149	0.5	11.5	109889	West
## New Hampshire	812	4281	0.7	3.3	9027	Northeast
## New Jersey	7333	5237	1.1	5.2	7521	Northeast
## New Mexico	1144	3601	2.2	9.7	121412	West
## New York	18076	4903	1.4	10.9	47831	Northeast
## North Carolina	5441	3875	1.8	11.1	48798	South
## North Dakota	637	5087	0.8	1.4	69273	North Central
## Ohio	10735	4561	0.8	7.4	40975	North Central
## Oklahoma	2715	3983	1.1	6.4	68782	South
## Oregon	2284	4660	0.6	4.2	96184	West
## Pennsylvania	11860	4449	1.0	6.1	44966	Northeast
## Rhode Island	931	4558	1.3	2.4	1049	Northeast
## South Carolina	2816	3635	2.3	11.6	30225	South
## South Dakota	681	4167	0.5	1.7	75955	North Central
## Tennessee	4173	3821	1.7	11.0	41328	South
## Texas	12237	4188	2.2	12.2	262134	South
## Utah	1203	4022	0.6	4.5	82096	West
## Vermont	472	3907	0.6	5.5	9267	Northeast
## Virginia	4981	4701	1.4	9.5	39780	South
## Washington	3559	4864	0.6	4.3	66570	West
## West Virginia	1799	3617	1.4	6.7	24070	South
## Wisconsin	4589	4468	0.7	3.0	54464	North Central
## Wyoming	376	4566	0.6	6.9	97203	West
##	categorizar.cut					
## Alabama	high					
## Alaska	some					
## Arizona	some					
## Arkansas	some					
## California	some					
## Colorado	low					
## Connecticut	some					
## Delaware	low					
## Florida	some					
## Georgia	some					
## Hawaii	some					
## Idaho	low					
## Illinois	low					
## Indiana	low					
## Iowa	low					
## Kansas	low					
## Kentucky	some					
## Louisiana	high					
## Maine	low					

```
## Maryland low
## Massachusetts some
## Michigan low
## Minnesota low
## Mississippi high
## Missouri low
## Montana low
## Nebraska low
## Nevada low
## New Hampshire low
## New Jersey some
## New Mexico high
## New York some
## North Carolina some
## North Dakota low
## Ohio low
## Oklahoma some
## Oregon low
## Pennsylvania low
## Rhode Island some
## South Carolina high
## South Dakota low
## Tennessee some
## Texas high
## Utah low
## Vermont low
## Virginia some
## Washington low
## West Virginia some
## Wisconsin low
## Wyoming low
```

#Usando ifelse

```
categorizar.ifelse = ifelse(state.x77.data.frame$Illiteracy >=0 &
                           state.x77.data.frame$Illiteracy < 1,
                           "low", ifelse(state.x77.data.frame$Illiteracy >=1
                                           & state.x77.data.frame$Illiteracy < 2, "some", "high"))
state.x77.data.frame = data.frame(state.x77.data.frame, categorizar.ifelse)
state.x77.data.frame
```

##	Population	Income	Illiteracy	Murder	Area	region
## Alabama	3615	3624	2.1	15.1	50708	South
## Alaska	365	6315	1.5	11.3	566432	West
## Arizona	2212	4530	1.8	7.8	113417	West
## Arkansas	2110	3378	1.9	10.1	51945	South
## California	21198	5114	1.1	10.3	156361	West
## Colorado	2541	4884	0.7	6.8	103766	West
## Connecticut	3100	5348	1.1	3.1	4862	Northeast
## Delaware	579	4809	0.9	6.2	1982	South
## Florida	8277	4815	1.3	10.7	54090	South
## Georgia	4931	4091	2.0	13.9	58073	South
## Hawaii	868	4963	1.9	6.2	6425	West
## Idaho	813	4119	0.6	5.3	82677	West
## Illinois	11197	5107	0.9	10.3	55748	North Central
## Indiana	5313	4458	0.7	7.1	36097	North Central

## Iowa	2861	4628	0.5	2.3	55941	North Central
## Kansas	2280	4669	0.6	4.5	81787	North Central
## Kentucky	3387	3712	1.6	10.6	39650	South
## Louisiana	3806	3545	2.8	13.2	44930	South
## Maine	1058	3694	0.7	2.7	30920	Northeast
## Maryland	4122	5299	0.9	8.5	9891	South
## Massachusetts	5814	4755	1.1	3.3	7826	Northeast
## Michigan	9111	4751	0.9	11.1	56817	North Central
## Minnesota	3921	4675	0.6	2.3	79289	North Central
## Mississippi	2341	3098	2.4	12.5	47296	South
## Missouri	4767	4254	0.8	9.3	68995	North Central
## Montana	746	4347	0.6	5.0	145587	West
## Nebraska	1544	4508	0.6	2.9	76483	North Central
## Nevada	590	5149	0.5	11.5	109889	West
## New Hampshire	812	4281	0.7	3.3	9027	Northeast
## New Jersey	7333	5237	1.1	5.2	7521	Northeast
## New Mexico	1144	3601	2.2	9.7	121412	West
## New York	18076	4903	1.4	10.9	47831	Northeast
## North Carolina	5441	3875	1.8	11.1	48798	South
## North Dakota	637	5087	0.8	1.4	69273	North Central
## Ohio	10735	4561	0.8	7.4	40975	North Central
## Oklahoma	2715	3983	1.1	6.4	68782	South
## Oregon	2284	4660	0.6	4.2	96184	West
## Pennsylvania	11860	4449	1.0	6.1	44966	Northeast
## Rhode Island	931	4558	1.3	2.4	1049	Northeast
## South Carolina	2816	3635	2.3	11.6	30225	South
## South Dakota	681	4167	0.5	1.7	75955	North Central
## Tennessee	4173	3821	1.7	11.0	41328	South
## Texas	12237	4188	2.2	12.2	262134	South
## Utah	1203	4022	0.6	4.5	82096	West
## Vermont	472	3907	0.6	5.5	9267	Northeast
## Virginia	4981	4701	1.4	9.5	39780	South
## Washington	3559	4864	0.6	4.3	66570	West
## West Virginia	1799	3617	1.4	6.7	24070	South
## Wisconsin	4589	4468	0.7	3.0	54464	North Central
## Wyoming	376	4566	0.6	6.9	97203	West
##	categorizar.cut		categorizar.ifelse			
## Alabama		high			high	
## Alaska		some			some	
## Arizona		some			some	
## Arkansas		some			some	
## California		some			some	
## Colorado		low			low	
## Connecticut		some			some	
## Delaware		low			low	
## Florida		some			some	
## Georgia		some			high	
## Hawaii		some			some	
## Idaho		low			low	
## Illinois		low			low	
## Indiana		low			low	
## Iowa		low			low	
## Kansas		low			low	
## Kentucky		some			some	

## Louisiana	high	high
## Maine	low	low
## Maryland	low	low
## Massachusetts	some	some
## Michigan	low	low
## Minnesota	low	low
## Mississippi	high	high
## Missouri	low	low
## Montana	low	low
## Nebraska	low	low
## Nevada	low	low
## New Hampshire	low	low
## New Jersey	some	some
## New Mexico	high	high
## New York	some	some
## North Carolina	some	some
## North Dakota	low	low
## Ohio	low	low
## Oklahoma	some	some
## Oregon	low	low
## Pennsylvania	low	some
## Rhode Island	some	some
## South Carolina	high	high
## South Dakota	low	low
## Tennessee	some	some
## Texas	high	high
## Utah	low	low
## Vermont	low	low
## Virginia	some	some
## Washington	low	low
## West Virginia	some	some
## Wisconsin	low	low
## Wyoming	low	low

- Encuentra que estado del oeste (west) tiene la formación mas baja y los mayores ingresos.
¿Que estado es?

```
datos1 = state.x77.data.frame [which(state.x77.data.frame$region == "West"),]
datos1[order(datos1$Illiteracy,order(datos1$Income, decreasing = TRUE)),][1,]
```

```
##      Population Income Illiteracy Murder   Area region categorizar.cut
## Nevada      590   5149      0.5   11.5 109889   West      low
##      categorizar.ifelse
## Nevada      low
```

* Crea un dataframe df with 40 columns, as follows:(df<-as.data.frame(matrix(sample(1:5, 2000, T), ncol=40)))

```
df <-as.data.frame(matrix(sample(1:5, 2000, T), ncol=40))
df
```

```
##      V1 V2 V3 V4 V5 V6 V7 V8 V9 V10 V11 V12 V13 V14 V15 V16 V17 V18 V19 V20
```

## 1	3	2	5	5	4	4	3	4	4	2	5	4	1	1	4	1	3	5	3	2
## 2	4	3	5	1	1	2	3	5	2	5	3	4	2	4	2	3	2	5	4	3
## 3	2	1	1	3	3	2	3	1	4	1	3	1	2	2	1	3	2	1	3	5
## 4	5	3	4	5	2	2	1	2	2	4	2	4	5	1	1	4	1	2	4	5
## 5	2	3	2	1	5	3	3	5	5	3	4	1	3	5	3	3	5	4	1	1
## 6	5	1	5	4	2	5	3	5	1	2	3	4	3	2	1	4	3	2	3	4
## 7	5	3	1	2	5	3	4	5	4	1	1	2	5	1	2	5	4	3	1	2
## 8	4	5	1	1	5	3	2	3	5	2	5	4	2	4	3	4	3	5	4	4
## 9	2	2	5	2	5	5	5	1	5	1	5	3	1	3	1	3	3	5	3	3
## 10	3	1	5	4	3	4	4	2	4	2	1	2	5	5	5	5	4	3	5	1
## 11	2	3	2	1	4	3	2	4	3	2	4	3	4	2	2	4	4	1	3	4
## 12	4	5	2	5	2	4	4	2	4	4	5	5	5	5	5	2	3	1	3	3
## 13	5	3	2	4	2	4	5	3	3	3	5	3	4	2	5	3	4	2	1	3
## 14	1	4	4	4	2	5	5	4	3	4	2	4	2	5	3	2	5	3	4	4
## 15	2	1	2	3	2	1	1	1	5	4	4	1	1	5	3	2	5	1	5	1
## 16	5	5	1	5	2	3	4	3	3	2	1	3	5	3	3	5	2	1	2	3
## 17	1	5	3	4	3	3	1	2	1	1	4	5	3	3	1	5	5	1	4	5
## 18	1	2	2	1	3	1	5	3	3	2	4	1	3	1	5	2	5	2	3	3
## 19	1	2	2	4	2	3	3	3	4	4	1	1	2	3	5	5	1	3	5	5
## 20	2	2	2	5	4	2	3	2	4	4	4	5	3	3	4	4	4	4	2	2
## 21	4	2	5	3	3	5	5	3	4	2	1	1	3	4	1	4	3	4	1	2
## 22	3	1	1	2	2	1	5	2	2	2	3	1	4	5	3	4	1	1	5	1
## 23	3	4	5	2	1	2	3	3	2	5	4	4	5	2	5	3	1	5	2	3
## 24	2	4	1	5	5	1	1	4	1	3	5	4	5	4	4	2	2	5	5	2
## 25	4	4	4	2	4	1	5	4	1	4	2	4	1	4	2	1	1	4	4	2
## 26	5	5	4	3	1	3	2	5	5	5	3	4	2	5	1	4	2	5	3	5
## 27	3	3	5	5	5	4	2	5	1	5	1	1	2	4	2	3	1	1	5	5
## 28	4	5	5	1	5	5	1	2	5	5	4	5	1	2	4	1	4	1	4	3
## 29	4	1	5	3	5	5	5	3	2	3	5	1	2	3	3	2	3	3	5	3
## 30	1	1	2	2	3	1	4	3	1	5	2	1	5	2	5	2	2	5	1	1
## 31	5	3	3	3	5	5	2	2	5	1	1	3	2	5	2	3	1	5	2	3
## 32	5	1	5	5	5	1	2	1	4	1	3	2	1	2	3	5	4	2	3	3
## 33	5	2	5	1	1	3	4	2	2	5	1	3	1	2	2	5	2	4	1	3
## 34	3	1	3	2	5	4	2	2	4	2	3	1	5	3	4	4	3	1	1	5
## 35	3	4	2	2	5	4	1	3	4	1	1	3	4	5	3	2	4	1	5	4
## 36	3	5	4	4	3	4	1	1	5	4	4	2	5	5	5	2	4	3	5	1
## 37	5	5	5	3	1	5	1	2	1	5	1	5	2	5	1	3	4	4	5	2
## 38	2	3	4	5	4	4	3	4	1	3	5	2	3	3	5	2	5	1	3	1
## 39	1	5	1	3	5	2	2	1	3	1	5	1	4	2	1	2	4	3	2	4
## 40	2	2	4	3	1	4	5	5	4	5	3	2	5	4	2	2	4	4	3	4
## 41	3	5	2	4	5	3	2	3	2	5	3	4	2	5	3	3	1	2	4	3
## 42	2	4	4	5	1	1	4	5	4	3	2	3	3	2	1	5	1	4	5	3
## 43	4	4	2	1	4	1	1	1	4	2	5	2	2	4	5	2	4	4	4	4
## 44	3	2	3	2	4	5	3	3	3	1	1	4	5	2	4	2	3	5	5	3
## 45	3	1	5	2	3	3	4	2	3	4	1	2	2	5	4	3	5	2	3	2
## 46	4	4	2	3	2	5	4	1	4	2	1	5	2	1	3	3	1	1	2	3
## 47	2	4	1	3	4	2	2	2	4	2	2	3	3	1	5	2	1	3	3	2
## 48	3	5	4	2	2	4	2	2	1	4	2	5	3	3	4	1	1	3	2	3
## 49	5	1	5	2	4	3	1	2	3	4	2	2	3	2	1	3	1	1	2	4
## 50	5	5	5	2	2	4	4	1	3	1	1	1	4	5	3	1	2	5	4	2
##	V21	V22	V23	V24	V25	V26	V27	V28	V29	V30	V31	V32	V33	V34	V35	V36	V37	V38		
## 1	5	3	4	2	3	3	3	3	4	5	3	4	4	4	4	1	3	5		
## 2	3	4	4	4	3	5	4	3	4	1	2	5	3	1	4	5	1	1		
## 3	3	2	3	3	2	1	5	2	3	2	3	3	4	2	2	1	5	3		

## 4	5	3	3	3	2	3	4	2	2	5	2	3	4	5	2	3	1	2
## 5	3	3	3	5	1	1	3	3	3	4	4	3	1	4	2	3	3	2
## 6	4	4	5	1	5	5	1	1	2	4	2	5	2	2	4	1	4	3
## 7	5	1	2	4	4	5	2	2	5	2	2	5	1	3	2	5	1	2
## 8	3	2	5	2	2	1	4	3	3	1	1	5	3	3	2	1	2	2
## 9	5	4	5	5	2	5	1	1	4	5	3	4	1	1	4	5	3	4
## 10	5	1	5	5	4	2	1	5	4	1	3	3	5	1	1	4	3	3
## 11	5	5	3	4	2	5	4	1	4	2	3	2	4	3	3	3	5	3
## 12	5	1	1	2	5	5	2	4	1	3	3	3	5	2	3	2	5	3
## 13	3	1	5	3	5	3	2	5	4	5	2	2	2	5	1	4	3	4
## 14	5	5	1	3	5	3	2	3	4	4	4	2	3	3	4	4	2	5
## 15	3	1	3	4	4	4	1	2	1	4	4	1	2	5	3	1	1	3
## 16	5	4	2	5	4	4	1	1	5	1	4	4	2	1	2	2	2	2
## 17	2	3	3	3	3	2	1	1	4	3	1	1	4	3	1	3	4	1
## 18	4	2	4	2	2	1	3	1	4	1	4	1	4	2	2	5	1	1
## 19	4	1	2	5	1	1	2	2	5	5	1	3	1	2	1	1	1	5
## 20	5	3	3	3	4	1	1	3	5	5	1	1	4	2	2	5	4	5
## 21	2	3	3	1	5	1	5	2	3	1	5	2	2	5	3	2	4	1
## 22	2	1	1	4	5	1	3	1	1	5	3	1	1	5	2	2	5	5
## 23	1	1	5	5	2	1	5	5	1	4	5	5	2	5	3	1	1	5
## 24	3	2	1	3	1	1	4	1	5	5	2	5	4	1	5	4	2	5
## 25	5	5	5	2	5	2	5	2	1	3	1	4	3	3	4	4	4	2
## 26	4	4	5	3	1	4	3	4	1	3	2	4	1	4	4	3	2	2
## 27	4	2	2	5	5	3	2	4	2	5	1	3	2	5	3	5	1	3
## 28	1	4	5	3	2	1	1	5	3	3	1	4	1	5	4	3	2	3
## 29	2	1	2	4	3	1	4	1	2	2	4	5	4	5	2	1	5	1
## 30	2	1	3	3	4	4	3	2	1	4	2	2	5	2	4	5	3	5
## 31	5	5	5	1	5	4	4	1	2	1	2	3	3	3	2	4	2	2
## 32	1	3	1	5	5	3	5	1	2	4	4	2	3	1	1	3	4	2
## 33	3	5	1	2	1	2	5	3	2	3	4	4	3	1	3	1	5	5
## 34	2	1	1	5	4	3	5	5	3	2	1	5	3	5	2	1	3	5
## 35	1	1	2	5	4	1	1	5	1	5	2	1	5	4	4	5	1	3
## 36	4	3	5	5	4	4	1	2	1	4	4	4	3	2	4	4	4	5
## 37	3	1	2	3	2	3	5	1	2	2	5	1	5	5	5	4	3	4
## 38	1	2	3	2	5	5	2	2	3	2	3	5	3	1	5	4	3	3
## 39	2	3	5	2	2	5	2	2	5	2	4	4	4	1	2	2	5	2
## 40	1	5	3	5	1	3	4	5	4	4	4	4	2	1	2	3	3	2
## 41	3	2	1	2	5	2	1	3	2	4	4	2	4	1	1	1	3	4
## 42	4	4	1	4	5	3	1	4	1	5	4	2	3	2	1	3	2	3
## 43	4	1	2	2	3	4	5	1	3	4	3	4	4	3	4	4	5	1
## 44	3	5	4	3	3	5	5	3	4	1	5	5	3	1	4	2	1	2
## 45	2	1	4	4	3	2	1	2	4	5	2	5	5	1	2	1	1	4
## 46	4	1	4	5	1	4	4	2	5	2	5	2	1	5	1	5	4	2
## 47	1	2	1	2	3	5	2	1	1	1	3	1	5	2	5	1	2	1
## 48	3	1	2	4	2	5	4	1	1	2	3	5	1	4	5	4	3	5
## 49	5	1	4	5	2	3	2	4	2	4	1	4	3	4	5	1	2	2
## 50	1	1	5	5	3	2	2	2	4	5	5	2	2	4	3	1	2	4
##	V39 V40																	
## 1	4	3																
## 2	4	3																
## 3	4	4																
## 4	3	4																
## 5	3	2																
## 6	2	5																

```
## 7    1    2
## 8    2    5
## 9    3    1
## 10   4    4
## 11   5    4
## 12   1    2
## 13   3    3
## 14   3    4
## 15   3    1
## 16   2    1
## 17   1    1
## 18   1    3
## 19   5    2
## 20   4    2
## 21   5    2
## 22   3    2
## 23   4    5
## 24   3    5
## 25   5    4
## 26   5    4
## 27   2    2
## 28   5    1
## 29   4    2
## 30   4    5
## 31   3    1
## 32   5    4
## 33   4    2
## 34   5    4
## 35   5    2
## 36   3    3
## 37   1    5
## 38   5    3
## 39   4    2
## 40   2    1
## 41   3    5
## 42   3    5
## 43   2    3
## 44   4    4
## 45   2    3
## 46   4    5
## 47   5    3
## 48   2    1
## 49   2    3
## 50   3    1
```

a. Ordena el dataframe por columnas, de izquierda a derecha en orden creciente

```
df <- df[do.call(order, df), ]
df
```

```
##      V1 V2 V3 V4 V5 V6 V7 V8 V9 V10 V11 V12 V13 V14 V15 V16 V17 V18 V19 V20
## 30   1  1  2  2  3  1  4  3  1  5  2  1  5  2  5  2  2  5  1  1
## 18   1  2  2  1  3  1  5  3  3  2  4  1  3  1  5  2  5  2  3  3
```

## 19	1	2	2	4	2	3	3	3	4	4	1	1	2	3	5	5	1	3	5	5
## 14	1	4	4	4	2	5	5	4	3	4	2	4	2	5	3	2	5	3	4	4
## 39	1	5	1	3	5	2	2	1	3	1	5	1	4	2	1	2	4	3	2	4
## 17	1	5	3	4	3	3	1	2	1	1	4	5	3	3	1	5	5	1	4	5
## 3	2	1	1	3	3	2	3	1	4	1	3	1	2	2	1	3	2	1	3	5
## 15	2	1	2	3	2	1	1	1	5	4	4	1	1	5	3	2	5	1	5	1
## 20	2	2	2	5	4	2	3	2	4	4	4	5	3	3	4	4	4	4	2	2
## 40	2	2	4	3	1	4	5	5	4	5	3	2	5	4	2	2	4	4	3	4
## 9	2	2	5	2	5	5	5	1	5	1	5	3	1	3	1	3	3	5	3	3
## 11	2	3	2	1	4	3	2	4	3	2	4	3	4	2	2	4	4	1	3	4
## 5	2	3	2	1	5	3	3	5	5	3	4	1	3	5	3	3	5	4	1	1
## 38	2	3	4	5	4	4	3	4	1	3	5	2	3	3	5	2	5	1	3	1
## 47	2	4	1	3	4	2	2	2	4	2	2	3	3	1	5	2	1	3	3	2
## 24	2	4	1	5	5	1	1	4	1	3	5	4	5	4	4	2	2	5	5	2
## 42	2	4	4	5	1	1	4	5	4	3	2	3	3	2	1	5	1	4	5	3
## 22	3	1	1	2	2	1	5	2	2	2	3	1	4	5	3	4	1	1	5	1
## 34	3	1	3	2	5	4	2	2	4	2	3	1	5	3	4	4	3	1	1	5
## 45	3	1	5	2	3	3	4	2	3	4	1	2	2	5	4	3	5	2	3	2
## 10	3	1	5	4	3	4	4	2	4	2	1	2	5	5	5	5	4	3	5	1
## 44	3	2	3	2	4	5	3	3	3	1	1	4	5	2	4	2	3	5	5	3
## 1	3	2	5	5	4	4	3	4	4	2	5	4	1	1	4	1	3	5	3	2
## 27	3	3	5	5	5	4	2	5	1	5	1	1	2	4	2	3	1	1	5	5
## 35	3	4	2	2	5	4	1	3	4	1	1	3	4	5	3	2	4	1	5	4
## 23	3	4	5	2	1	2	3	3	2	5	4	4	5	2	5	3	1	5	2	3
## 41	3	5	2	4	5	3	2	3	2	5	3	4	2	5	3	3	1	2	4	3
## 48	3	5	4	2	2	4	2	2	1	4	2	5	3	3	4	1	1	3	2	3
## 36	3	5	4	4	3	4	1	1	5	4	4	2	5	5	5	2	4	3	5	1
## 29	4	1	5	3	5	5	5	3	2	3	5	1	2	3	3	2	3	3	5	3
## 21	4	2	5	3	3	5	5	3	4	2	1	1	3	4	1	4	3	4	1	2
## 2	4	3	5	1	1	2	3	5	2	5	3	4	2	4	2	3	2	5	4	3
## 43	4	4	2	1	4	1	1	1	4	2	5	2	2	4	5	2	4	4	4	4
## 46	4	4	2	3	2	5	4	1	4	2	1	5	2	1	3	3	1	1	2	3
## 25	4	4	4	2	4	1	5	4	1	4	2	4	1	4	2	1	1	4	4	2
## 8	4	5	1	1	5	3	2	3	5	2	5	4	2	4	3	4	3	5	4	4
## 12	4	5	2	5	2	4	4	2	4	4	5	5	5	5	5	2	3	1	3	3
## 28	4	5	5	1	5	5	1	2	5	5	4	5	1	2	4	1	4	1	4	3
## 49	5	1	5	2	4	3	1	2	3	4	2	2	3	2	1	3	1	1	2	4
## 6	5	1	5	4	2	5	3	5	1	2	3	4	3	2	1	4	3	2	3	4
## 32	5	1	5	5	5	1	2	1	4	1	3	2	1	2	3	5	4	2	3	3
## 33	5	2	5	1	1	3	4	2	2	5	1	3	1	2	2	5	2	4	1	3
## 7	5	3	1	2	5	3	4	5	4	1	1	2	5	1	2	5	4	3	1	2
## 13	5	3	2	4	2	4	5	3	3	3	5	3	4	2	5	3	4	2	1	3
## 31	5	3	3	3	5	5	2	2	5	1	1	3	2	5	2	3	1	5	2	3
## 4	5	3	4	5	2	2	1	2	2	4	2	4	5	1	1	4	1	2	4	5
## 16	5	5	1	5	2	3	4	3	3	2	1	3	5	3	3	5	2	1	2	3
## 26	5	5	4	3	1	3	2	5	5	5	3	4	2	5	1	4	2	5	3	5
## 50	5	5	5	2	2	4	4	1	3	1	1	1	4	5	3	1	2	5	4	2
## 37	5	5	5	3	1	5	1	2	1	5	1	5	2	5	1	3	4	4	5	2
##	V21	V22	V23	V24	V25	V26	V27	V28	V29	V30	V31	V32	V33	V34	V35	V36	V37	V38		
## 30	2	1	3	3	4	4	3	2	1	4	2	2	5	2	4	5	3	5		
## 18	4	2	4	2	2	1	3	1	4	1	4	1	4	2	2	5	1	1		
## 19	4	1	2	5	1	1	2	2	5	5	1	3	1	2	1	1	1	5		
## 14	5	5	1	3	5	3	2	3	4	4	4	2	3	3	4	4	2	5		
## 39	2	3	5	2	2	5	2	2	5	2	4	4	4	1	2	2	5	2		

## 17	2	3	3	3	3	2	1	1	4	3	1	1	4	3	1	3	4	1
## 3	3	2	3	3	2	1	5	2	3	2	3	3	4	2	2	1	5	3
## 15	3	1	3	4	4	4	1	2	1	4	4	1	2	5	3	1	1	3
## 20	5	3	3	3	4	1	1	3	5	5	1	1	4	2	2	5	4	5
## 40	1	5	3	5	1	3	4	5	4	4	4	4	2	1	2	3	3	2
## 9	5	4	5	5	2	5	1	1	4	5	3	4	1	1	4	5	3	4
## 11	5	5	3	4	2	5	4	1	4	2	3	2	4	3	3	3	5	3
## 5	3	3	3	5	1	1	3	3	3	4	4	3	1	4	2	3	3	2
## 38	1	2	3	2	5	5	2	2	3	2	3	5	3	1	5	4	3	3
## 47	1	2	1	2	3	5	2	1	1	1	3	1	5	2	5	1	2	1
## 24	3	2	1	3	1	1	4	1	5	5	2	5	4	1	5	4	2	5
## 42	4	4	1	4	5	3	1	4	1	5	4	2	3	2	1	3	2	3
## 22	2	1	1	4	5	1	3	1	1	5	3	1	1	5	2	2	5	5
## 34	2	1	1	5	4	3	5	5	3	2	1	5	3	5	2	1	3	5
## 45	2	1	4	4	3	2	1	2	4	5	2	5	5	1	2	1	1	4
## 10	5	1	5	5	4	2	1	5	4	1	3	3	5	1	1	4	3	3
## 44	3	5	4	3	3	5	5	3	4	1	5	5	3	1	4	2	1	2
## 1	5	3	4	2	3	3	3	3	4	5	3	4	4	4	4	1	3	5
## 27	4	2	2	5	5	3	2	4	2	5	1	3	2	5	3	5	1	3
## 35	1	1	2	5	4	1	1	5	1	5	2	1	5	4	4	5	1	3
## 23	1	1	5	5	2	1	5	5	1	4	5	5	2	5	3	1	1	5
## 41	3	2	1	2	5	2	1	3	2	4	4	2	4	1	1	1	3	4
## 48	3	1	2	4	2	5	4	1	1	2	3	5	1	4	5	4	3	5
## 36	4	3	5	5	4	4	1	2	1	4	4	4	3	2	4	4	4	5
## 29	2	1	2	4	3	1	4	1	2	2	4	5	4	5	2	1	5	1
## 21	2	3	3	1	5	1	5	2	3	1	5	2	2	5	3	2	4	1
## 2	3	4	4	4	3	5	4	3	4	1	2	5	3	1	4	5	1	1
## 43	4	1	2	2	3	4	5	1	3	4	3	4	4	3	4	4	5	1
## 46	4	1	4	5	1	4	4	2	5	2	5	2	1	5	1	5	4	2
## 25	5	5	5	2	5	2	5	2	1	3	1	4	3	3	4	4	4	2
## 8	3	2	5	2	2	1	4	3	3	1	1	5	3	3	2	1	2	2
## 12	5	1	1	2	5	5	2	4	1	3	3	3	5	2	3	2	5	3
## 28	1	4	5	3	2	1	1	5	3	3	1	4	1	5	4	3	2	3
## 49	5	1	4	5	2	3	2	4	2	4	1	4	3	4	5	1	2	2
## 6	4	4	5	1	5	5	1	1	2	4	2	5	2	2	4	1	4	3
## 32	1	3	1	5	5	3	5	1	2	4	4	2	3	1	1	3	4	2
## 33	3	5	1	2	1	2	5	3	2	3	4	4	3	1	3	1	5	5
## 7	5	1	2	4	4	5	2	2	5	2	2	5	1	3	2	5	1	2
## 13	3	1	5	3	5	3	2	5	4	5	2	2	2	5	1	4	3	4
## 31	5	5	5	1	5	4	4	1	2	1	2	3	3	3	2	4	2	2
## 4	5	3	3	3	2	3	4	2	2	5	2	3	4	5	2	3	1	2
## 16	5	4	2	5	4	4	1	1	5	1	4	4	2	1	2	2	2	2
## 26	4	4	5	3	1	4	3	4	1	3	2	4	1	4	4	3	2	2
## 50	1	1	5	5	3	2	2	2	4	5	5	2	2	4	3	1	2	4
## 37	3	1	2	3	2	3	5	1	2	2	5	1	5	5	5	4	3	4
##	V39	V40																
## 30	4	5																
## 18	1	3																
## 19	5	2																
## 14	3	4																
## 39	4	2																
## 17	1	1																
## 3	4	4																
## 15	3	1																

```
## 20  4  2
## 40  2  1
## 9   3  1
## 11  5  4
## 5   3  2
## 38  5  3
## 47  5  3
## 24  3  5
## 42  3  5
## 22  3  2
## 34  5  4
## 45  2  3
## 10  4  4
## 44  4  4
## 1   4  3
## 27  2  2
## 35  5  2
## 23  4  5
## 41  3  5
## 48  2  1
## 36  3  3
## 29  4  2
## 21  5  2
## 2   4  3
## 43  2  3
## 46  4  5
## 25  5  4
## 8   2  5
## 12  1  2
## 28  5  1
## 49  2  3
## 6   2  5
## 32  5  4
## 33  4  2
## 7   1  2
## 13  3  3
## 31  3  1
## 4   3  4
## 16  2  1
## 26  5  4
## 50  3  1
## 37  1  5
```

b. Ordena el dataframe por columnas, de izquierda a derecha en orden decreciente

```
df <- df[do.call(order, -df), ]
```

```
df
```

```
##      V1 V2 V3 V4 V5 V6 V7 V8 V9 V10 V11 V12 V13 V14 V15 V16 V17 V18 V19 V20
## 37  5  5  5  3  1  5  1  2  1  5  1  5  2  5  1  3  4  4  5  2
## 50  5  5  5  2  2  4  4  1  3  1  1  1  4  5  3  1  2  5  4  2
## 26  5  5  4  3  1  3  2  5  5  5  3  4  2  5  1  4  2  5  3  5
```


##	16	5	5	1	5	2	3	4	3	3	2	1	3	5	3	3	5	2	1	2	3
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##	31	5	3	3	3	5	5	2	2	5	1	1	3	2	5	2	3	1	5	2	3
##	13	5	3	2	4	2	4	5	3	3	3	5	3	4	2	5	3	4	2	1	3
##	7	5	3	1	2	5	3	4	5	4	1	1	2	5	1	2	5	4	3	1	2
##	33	5	2	5	1	1	3	4	2	2	5	1	3	1	2	2	5	2	4	1	3
##	32	5	1	5	5	5	1	2	1	4	1	3	2	1	2	3	5	4	2	3	3
##	6	5	1	5	4	2	5	3	5	1	2	3	4	3	2	1	4	3	2	3	4
##	49	5	1	5	2	4	3	1	2	3	4	2	2	3	2	1	3	1	1	2	4
##	28	4	5	5	1	5	5	1	2	5	5	4	5	1	2	4	1	4	1	4	3
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##	38	2	3	4	5	4	4	3	4	1	3	5	2	3	3	5	2	5	1	3	1
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##	9	2	2	5	2	5	5	5	1	5	1	5	3	1	3	1	3	3	5	3	3
##	40	2	2	4	3	1	4	5	5	4	5	3	2	5	4	2	2	4	4	3	4
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##	15	2	1	2	3	2	1	1	1	5	4	4	1	1	5	3	2	5	1	5	1
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##	17	1	5	3	4	3	3	1	2	1	1	4	5	3	3	1	5	5	1	4	5
##	39	1	5	1	3	5	2	2	1	3	1	5	1	4	2	1	2	4	3	2	4
##	14	1	4	4	4	2	5	5	4	3	4	2	4	2	5	3	2	5	3	4	4
##	19	1	2	2	4	2	3	3	3	4	4	1	1	2	3	5	5	1	3	5	5
##	18	1	2	2	1	3	1	5	3	3	2	4	1	3	1	5	2	5	2	3	3
##	30	1	1	2	2	3	1	4	3	1	5	2	1	5	2	5	2	2	5	1	1
##		V21	V22	V23	V24	V25	V26	V27	V28	V29	V30	V31	V32	V33	V34	V35	V36	V37	V38		
##	37	3	1	2	3	2	3	5	1	2	2	5	1	5	5	5	5	4	3	4	
##	50	1	1	5	5	3	2	2	2	4	5	5	2	2	4	3	1	2	4		
##	26	4	4	5	3	1	4	3	4	1	3	2	4	1	4	4	3	2	2		
##	16	5	4	2	5	4	4	1	1	5	1	4	4	2	1	2	2	2	2		
##	4	5	3	3	3	2	3	4	2	2	5	2	3	4	5	2	3	1	2		
##	31	5	5	5	1	5	4	4	1	2	1	2	3	3	3	2	4	2	2		

## 13	3	1	5	3	5	3	2	5	4	5	2	2	2	5	1	4	3	4
## 7	5	1	2	4	4	5	2	2	5	2	2	5	1	3	2	5	1	2
## 33	3	5	1	2	1	2	5	3	2	3	4	4	3	1	3	1	5	5
## 32	1	3	1	5	5	3	5	1	2	4	4	2	3	1	1	3	4	2
## 6	4	4	5	1	5	5	1	1	2	4	2	5	2	2	4	1	4	3
## 49	5	1	4	5	2	3	2	4	2	4	1	4	3	4	5	1	2	2
## 28	1	4	5	3	2	1	1	5	3	3	1	4	1	5	4	3	2	3
## 12	5	1	1	2	5	5	2	4	1	3	3	3	5	2	3	2	5	3
## 8	3	2	5	2	2	1	4	3	3	1	1	5	3	3	2	1	2	2
## 25	5	5	5	2	5	2	5	2	1	3	1	4	3	3	4	4	4	2
## 46	4	1	4	5	1	4	4	2	5	2	5	2	1	5	1	5	4	2
## 43	4	1	2	2	3	4	5	1	3	4	3	4	4	3	4	4	5	1
## 2	3	4	4	4	3	5	4	3	4	1	2	5	3	1	4	5	1	1
## 21	2	3	3	1	5	1	5	2	3	1	5	2	2	5	3	2	4	1
## 29	2	1	2	4	3	1	4	1	2	2	4	5	4	5	2	1	5	1
## 36	4	3	5	5	4	4	1	2	1	4	4	4	3	2	4	4	4	5
## 48	3	1	2	4	2	5	4	1	1	2	3	5	1	4	5	4	3	5
## 41	3	2	1	2	5	2	1	3	2	4	4	2	4	1	1	1	3	4
## 23	1	1	5	5	2	1	5	5	1	4	5	5	2	5	3	1	1	5
## 35	1	1	2	5	4	1	1	5	1	5	2	1	5	4	4	5	1	3
## 27	4	2	2	5	5	3	2	4	2	5	1	3	2	5	3	5	1	3
## 1	5	3	4	2	3	3	3	3	4	5	3	4	4	4	4	1	3	5
## 44	3	5	4	3	3	5	5	3	4	1	5	5	3	1	4	2	1	2
## 10	5	1	5	5	4	2	1	5	4	1	3	3	5	1	1	4	3	3
## 45	2	1	4	4	3	2	1	2	4	5	2	5	5	1	2	1	1	4
## 34	2	1	1	5	4	3	5	5	3	2	1	5	3	5	2	1	3	5
## 22	2	1	1	4	5	1	3	1	1	5	3	1	1	5	2	2	5	5
## 42	4	4	1	4	5	3	1	4	1	5	4	2	3	2	1	3	2	3
## 24	3	2	1	3	1	1	4	1	5	5	2	5	4	1	5	4	2	5
## 47	1	2	1	2	3	5	2	1	1	1	3	1	5	2	5	1	2	1
## 38	1	2	3	2	5	5	2	2	3	2	3	5	3	1	5	4	3	3
## 5	3	3	3	5	1	1	3	3	3	4	4	3	1	4	2	3	3	2
## 11	5	5	3	4	2	5	4	1	4	2	3	2	4	3	3	3	5	3
## 9	5	4	5	5	2	5	1	1	4	5	3	4	1	1	4	5	3	4
## 40	1	5	3	5	1	3	4	5	4	4	4	4	2	1	2	3	3	2
## 20	5	3	3	3	4	1	1	3	5	5	1	1	4	2	2	5	4	5
## 15	3	1	3	4	4	4	1	2	1	4	4	1	2	5	3	1	1	3
## 3	3	2	3	3	2	1	5	2	3	2	3	3	4	2	2	1	5	3
## 17	2	3	3	3	3	2	1	1	4	3	1	1	4	3	1	3	4	1
## 39	2	3	5	2	2	5	2	2	5	2	4	4	4	1	2	2	5	2
## 14	5	5	1	3	5	3	2	3	4	4	4	2	3	3	4	4	2	5
## 19	4	1	2	5	1	1	2	2	5	5	1	3	1	2	1	1	1	5
## 18	4	2	4	2	2	1	3	1	4	1	4	1	4	2	2	5	1	1
## 30	2	1	3	3	4	4	3	2	1	4	2	2	5	2	4	5	3	5
##	V39	V40																
## 37	1	5																
## 50	3	1																
## 26	5	4																
## 16	2	1																
## 4	3	4																
## 31	3	1																
## 13	3	3																
## 7	1	2																
## 33	4	2																

```
## 32  5  4
##  6  2  5
## 49  2  3
## 28  5  1
## 12  1  2
##  8  2  5
## 25  5  4
## 46  4  5
## 43  2  3
##  2  4  3
## 21  5  2
## 29  4  2
## 36  3  3
## 48  2  1
## 41  3  5
## 23  4  5
## 35  5  2
## 27  2  2
##  1  4  3
## 44  4  4
## 10  4  4
## 45  2  3
## 34  5  4
## 22  3  2
## 42  3  5
## 24  3  5
## 47  5  3
## 38  5  3
##  5  3  2
## 11  5  4
##  9  3  1
## 40  2  1
## 20  4  2
## 15  3  1
##  3  4  4
## 17  1  1
## 39  4  2
## 14  3  4
## 19  5  2
## 18  1  3
## 30  4  5
```

c. Ordena el dataframe por columnas, de derecha a izquierda en orden creciente

```
df <- df[do.call(order, rev(df)), ]
df
```

```
##      V1 V2 V3 V4 V5 V6 V7 V8 V9 V10 V11 V12 V13 V14 V15 V16 V17 V18 V19 V20
## 17  1  5  3  4  3  3  1  2  1  1  4  5  3  3  1  5  5  1  4  5
## 16  5  5  1  5  2  3  4  3  3  2  1  3  5  3  3  5  2  1  2  3
## 40  2  2  4  3  1  4  5  5  4  5  3  2  5  4  2  2  4  4  3  4
## 48  3  5  4  2  2  4  2  2  1  4  2  5  3  3  4  1  1  3  2  3
## 31  5  3  3  3  5  5  2  2  5  1  1  3  2  5  2  3  1  5  2  3
```

## 15	2	1	2	3	2	1	1	1	5	4	4	1	1	5	3	2	5	1	5	1
## 50	5	5	5	2	2	4	4	1	3	1	1	1	4	5	3	1	2	5	4	2
## 9	2	2	5	2	5	5	5	1	5	1	5	3	1	3	1	3	3	5	3	3
## 28	4	5	5	1	5	5	1	2	5	5	4	5	1	2	4	1	4	1	4	3
## 7	5	3	1	2	5	3	4	5	4	1	1	2	5	1	2	5	4	3	1	2
## 12	4	5	2	5	2	4	4	2	4	4	5	5	5	5	5	2	3	1	3	3
## 27	3	3	5	5	5	4	2	5	1	5	1	1	2	4	2	3	1	1	5	5
## 5	2	3	2	1	5	3	3	5	5	3	4	1	3	5	3	3	5	4	1	1
## 22	3	1	1	2	2	1	5	2	2	2	3	1	4	5	3	4	1	1	5	1
## 29	4	1	5	3	5	5	5	3	2	3	5	1	2	3	3	2	3	3	5	3
## 39	1	5	1	3	5	2	2	1	3	1	5	1	4	2	1	2	4	3	2	4
## 20	2	2	2	5	4	2	3	2	4	4	4	5	3	3	4	4	4	4	2	2
## 33	5	2	5	1	1	3	4	2	2	5	1	3	1	2	2	5	2	4	1	3
## 21	4	2	5	3	3	5	5	3	4	2	1	1	3	4	1	4	3	4	1	2
## 35	3	4	2	2	5	4	1	3	4	1	1	3	4	5	3	2	4	1	5	4
## 19	1	2	2	4	2	3	3	3	4	4	1	1	2	3	5	5	1	3	5	5
## 18	1	2	2	1	3	1	5	3	3	2	4	1	3	1	5	2	5	2	3	3
## 43	4	4	2	1	4	1	1	1	4	2	5	2	2	4	5	2	4	4	4	4
## 49	5	1	5	2	4	3	1	2	3	4	2	2	3	2	1	3	1	1	2	4
## 45	3	1	5	2	3	3	4	2	3	4	1	2	2	5	4	3	5	2	3	2
## 13	5	3	2	4	2	4	5	3	3	3	5	3	4	2	5	3	4	2	1	3
## 36	3	5	4	4	3	4	1	1	5	4	4	2	5	5	5	2	4	3	5	1
## 2	4	3	5	1	1	2	3	5	2	5	3	4	2	4	2	3	2	5	4	3
## 1	3	2	5	5	4	4	3	4	4	2	5	4	1	1	4	1	3	5	3	2
## 47	2	4	1	3	4	2	2	2	4	2	2	3	3	1	5	2	1	3	3	2
## 38	2	3	4	5	4	4	3	4	1	3	5	2	3	3	5	2	5	1	3	1
## 4	5	3	4	5	2	2	1	2	2	4	2	4	5	1	1	4	1	2	4	5
## 14	1	4	4	4	2	5	5	4	3	4	2	4	2	5	3	2	5	3	4	4
## 44	3	2	3	2	4	5	3	3	3	1	1	4	5	2	4	2	3	5	5	3
## 10	3	1	5	4	3	4	4	2	4	2	1	2	5	5	5	5	4	3	5	1
## 3	2	1	1	3	3	2	3	1	4	1	3	1	2	2	1	3	2	1	3	5
## 26	5	5	4	3	1	3	2	5	5	5	3	4	2	5	1	4	2	5	3	5
## 32	5	1	5	5	5	1	2	1	4	1	3	2	1	2	3	5	4	2	3	3
## 25	4	4	4	2	4	1	5	4	1	4	2	4	1	4	2	1	1	4	4	2
## 11	2	3	2	1	4	3	2	4	3	2	4	3	4	2	2	4	4	1	3	4
## 34	3	1	3	2	5	4	2	2	4	2	3	1	5	3	4	4	3	1	1	5
## 37	5	5	5	3	1	5	1	2	1	5	1	5	2	5	1	3	4	4	5	2
## 8	4	5	1	1	5	3	2	3	5	2	5	4	2	4	3	4	3	5	4	4
## 6	5	1	5	4	2	5	3	5	1	2	3	4	3	2	1	4	3	2	3	4
## 42	2	4	4	5	1	1	4	5	4	3	2	3	3	2	1	5	1	4	5	3
## 41	3	5	2	4	5	3	2	3	2	5	3	4	2	5	3	3	1	2	4	3
## 24	2	4	1	5	5	1	1	4	1	3	5	4	5	4	4	2	2	5	5	2
## 46	4	4	2	3	2	5	4	1	4	2	1	5	2	1	3	3	1	1	2	3
## 23	3	4	5	2	1	2	3	3	2	5	4	4	5	2	5	3	1	5	2	3
## 30	1	1	2	2	3	1	4	3	1	5	2	1	5	2	5	2	2	5	1	1
##	V21	V22	V23	V24	V25	V26	V27	V28	V29	V30	V31	V32	V33	V34	V35	V36	V37	V38		
## 17	2	3	3	3	3	2	1	1	4	3	1	1	4	3	1	3	4	1		
## 16	5	4	2	5	4	4	1	1	5	1	4	4	2	1	2	2	2	2		
## 40	1	5	3	5	1	3	4	5	4	4	4	4	2	1	2	3	3	2		
## 48	3	1	2	4	2	5	4	1	1	2	3	5	1	4	5	4	3	5		
## 31	5	5	5	1	5	4	4	1	2	1	2	3	3	3	2	4	2	2		
## 15	3	1	3	4	4	4	1	2	1	4	4	1	2	5	3	1	1	3		
## 50	1	1	5	5	3	2	2	2	4	5	5	2	2	4	3	1	2	4		
## 9	5	4	5	5	2	5	1	1	4	5	3	4	1	1	4	5	3	4		

## 28	1	4	5	3	2	1	1	5	3	3	1	4	1	5	4	3	2	3
## 7	5	1	2	4	4	5	2	2	5	2	2	5	1	3	2	5	1	2
## 12	5	1	1	2	5	5	2	4	1	3	3	3	5	2	3	2	5	3
## 27	4	2	2	5	5	3	2	4	2	5	1	3	2	5	3	5	1	3
## 5	3	3	3	5	1	1	3	3	3	4	4	3	1	4	2	3	3	2
## 22	2	1	1	4	5	1	3	1	1	5	3	1	1	5	2	2	5	5
## 29	2	1	2	4	3	1	4	1	2	2	4	5	4	5	2	1	5	1
## 39	2	3	5	2	2	5	2	2	5	2	4	4	4	1	2	2	5	2
## 20	5	3	3	3	4	1	1	3	5	5	1	1	4	2	2	5	4	5
## 33	3	5	1	2	1	2	5	3	2	3	4	4	3	1	3	1	5	5
## 21	2	3	3	1	5	1	5	2	3	1	5	2	2	5	3	2	4	1
## 35	1	1	2	5	4	1	1	5	1	5	2	1	5	4	4	5	1	3
## 19	4	1	2	5	1	1	2	2	5	5	1	3	1	2	1	1	1	5
## 18	4	2	4	2	2	1	3	1	4	1	4	1	4	2	2	5	1	1
## 43	4	1	2	2	3	4	5	1	3	4	3	4	4	3	4	4	5	1
## 49	5	1	4	5	2	3	2	4	2	4	1	4	3	4	5	1	2	2
## 45	2	1	4	4	3	2	1	2	4	5	2	5	5	1	2	1	1	4
## 13	3	1	5	3	5	3	2	5	4	5	2	2	2	5	1	4	3	4
## 36	4	3	5	5	4	4	1	2	1	4	4	4	3	2	4	4	4	5
## 2	3	4	4	4	3	5	4	3	4	1	2	5	3	1	4	5	1	1
## 1	5	3	4	2	3	3	3	3	4	5	3	4	4	4	4	1	3	5
## 47	1	2	1	2	3	5	2	1	1	1	3	1	5	2	5	1	2	1
## 38	1	2	3	2	5	5	2	2	3	2	3	5	3	1	5	4	3	3
## 4	5	3	3	3	2	3	4	2	2	5	2	3	4	5	2	3	1	2
## 14	5	5	1	3	5	3	2	3	4	4	4	2	3	3	4	4	2	5
## 44	3	5	4	3	3	5	5	3	4	1	5	5	3	1	4	2	1	2
## 10	5	1	5	5	4	2	1	5	4	1	3	3	5	1	1	4	3	3
## 3	3	2	3	3	2	1	5	2	3	2	3	3	4	2	2	1	5	3
## 26	4	4	5	3	1	4	3	4	1	3	2	4	1	4	4	3	2	2
## 32	1	3	1	5	5	3	5	1	2	4	4	2	3	1	1	3	4	2
## 25	5	5	5	2	5	2	5	2	1	3	1	4	3	3	4	4	4	2
## 11	5	5	3	4	2	5	4	1	4	2	3	2	4	3	3	3	5	3
## 34	2	1	1	5	4	3	5	5	3	2	1	5	3	5	2	1	3	5
## 37	3	1	2	3	2	3	5	1	2	2	5	1	5	5	5	4	3	4
## 8	3	2	5	2	2	1	4	3	3	1	1	5	3	3	2	1	2	2
## 6	4	4	5	1	5	5	1	1	2	4	2	5	2	2	4	1	4	3
## 42	4	4	1	4	5	3	1	4	1	5	4	2	3	2	1	3	2	3
## 41	3	2	1	2	5	2	1	3	2	4	4	2	4	1	1	1	3	4
## 24	3	2	1	3	1	1	4	1	5	5	2	5	4	1	5	4	2	5
## 46	4	1	4	5	1	4	4	2	5	2	5	2	1	5	1	5	4	2
## 23	1	1	5	5	2	1	5	5	1	4	5	5	2	5	3	1	1	5
## 30	2	1	3	3	4	4	3	2	1	4	2	2	5	2	4	5	3	5
##	V39	V40																
## 17	1	1																
## 16	2	1																
## 40	2	1																
## 48	2	1																
## 31	3	1																
## 15	3	1																
## 50	3	1																
## 9	3	1																
## 28	5	1																
## 7	1	2																
## 12	1	2																

```
## 27  2  2
## 5   3  2
## 22  3  2
## 29  4  2
## 39  4  2
## 20  4  2
## 33  4  2
## 21  5  2
## 35  5  2
## 19  5  2
## 18  1  3
## 43  2  3
## 49  2  3
## 45  2  3
## 13  3  3
## 36  3  3
## 2   4  3
## 1   4  3
## 47  5  3
## 38  5  3
## 4   3  4
## 14  3  4
## 44  4  4
## 10  4  4
## 3   4  4
## 26  5  4
## 32  5  4
## 25  5  4
## 11  5  4
## 34  5  4
## 37  1  5
## 8   2  5
## 6   2  5
## 42  3  5
## 41  3  5
## 24  3  5
## 46  4  5
## 23  4  5
## 30  4  5
```

2. Importando información

* Vamos a trabajar con otro dataframe. Descarga el fichero student.txt de la plataforma PRADO, almacena la información en una variable llamada “students”. Ten en cuenta que los datos son tab-delimited y tienen un texto para cada columna. Comprueba que R ha leído correctamente el fichero imprimiendo el objeto en la pantalla

```
students = read.delim(file = "student.txt")
students
```

```
##   height shoesize gender population
## 1    181      44   male     kuopio
## 2    160      38 female     kuopio
```

```
## 3      174      42 female    kuopio
## 4      170      43   male    kuopio
## 5      172      43   male    kuopio
## 6      165      39 female    kuopio
## 7      161      38 female    kuopio
## 8      167      38 female    tampere
## 9      164      39 female    tampere
## 10     166      38 female    tampere
## 11     162      37 female    tampere
## 12     158      36 female    tampere
## 13     175      42   male    tampere
## 14     181      44   male    tampere
## 15     180      43   male    tampere
## 16     177      43   male    tampere
## 17     173      41   male    tampere
```

Imprime solo los nombres de la columnas

```
colnames(students)
```

```
## [1] "height"      "shoesize"    "gender"      "population"
```

Llama a la columna height solo

```
students$height
```

```
## [1] 181 160 174 170 172 165 161 167 164 166 162 158 175 181 180 177 173
```

¿Cuántas observaciones hay en cada grupo?. Utiliza la función `table()`. Este commando se puede utilizar para crear tablas cruzadas (cross-tabulations)

```
table(students$gender)
```

```
##
## female    male
##         9      8
```

```
table(students$gender, students$height)
```

```
##
##          158 160 161 162 164 165 166 167 170 172 173 174 175 177 180 181
## female    1  1  1  1  1  1  1  1  0  0  0  1  0  0  0  0
## male      0  0  0  0  0  0  0  0  1  1  1  0  1  1  1  2
```

Crea nuevas variables a partir de los datos que tenemos. Vamos a crear una variable nueva “sym” que contenga M si el genero es masculino y F si el genero es femenino. Busca en la ayuda información sobre la función `ifelse()`. Crea una segunda variable “colours” cuyo valor ser?? “Blue” si el estudiante es de kuopio y “Red” si es de otro sitio.

```
students$gender
```

```
## [1] male   female female male   male   female female female female female
## [11] female female male   male   male   male   male
## Levels: female male
```

```
sym = ifelse(students$gender == "male", "M", "F")
sym
```

```
## [1] "M" "F" "F" "M" "M" "F" "F" "F" "F" "F" "F" "F" "M" "M" "M" "M" "M"
```

```
colours = ifelse(students$population == "kuopio", "Blue", "Red")
colours
```

```
## [1] "Blue" "Blue" "Blue" "Blue" "Blue" "Blue" "Blue" "Blue" "Red"  "Red"  "Red"
## [11] "Red"  "Red"  "Red"  "Red"  "Red"  "Red"  "Red"  "Red"
```

Con los datos anteriores de height y shoesize y las nuevas variables crea un nuevo data.frame que se llame students.new

```
students.new = data.frame(students$height, students$shoesize, sym, colours)
students.new
```

```
##      students.height students.shoesize sym colours
## 1             181             44  M    Blue
## 2             160             38  F    Blue
## 3             174             42  F    Blue
## 4             170             43  M    Blue
## 5             172             43  M    Blue
## 6             165             39  F    Blue
## 7             161             38  F    Blue
## 8             167             38  F    Red
## 9             164             39  F    Red
## 10            166             38  F    Red
## 11            162             37  F    Red
## 12            158             36  F    Red
## 13            175             42  M    Red
## 14            181             44  M    Red
## 15            180             43  M    Red
## 16            177             43  M    Red
## 17            173             41  M    Red
```

Comprueba que la clase de student.new es un dataframe

```
class(students.new)
```

```
## [1] "data.frame"
```

Crea dos subsets a partir del dataset student. Dividelo dependiendo del sexo. Para ello primero comprueba que estudiantes son hombres (male). Pista: busca información sobre la función which.

```
which(students$gender == "male")
```

```
## [1] 1 4 5 13 14 15 16 17
```


Basándote en esa selección dada por `which()` toma solo esas filas del dataset `student` para generar el subset `student.male`

- Repite el procedimiento para seleccionar las estudiantes mujeres (females)

```
students.male = students[which(students$gender == "male"),]
students.male

##    height shoesize gender population
## 1     181       44   male      kuopio
## 4     170       43   male      kuopio
## 5     172       43   male      kuopio
## 13    175       42   male    tampere
## 14    181       44   male    tampere
## 15    180       43   male    tampere
## 16    177       43   male    tampere
## 17    173       41   male    tampere

students.female = students[which(students$gender == "female"),]
students.female
```

```
##    height shoesize gender population
## 2     160       38 female      kuopio
## 3     174       42 female      kuopio
## 6     165       39 female      kuopio
## 7     161       38 female      kuopio
## 8     167       38 female    tampere
## 9     164       39 female    tampere
## 10    166       38 female    tampere
## 11    162       37 female    tampere
## 12    158       36 female    tampere
```

Utiliza la function `write.table()` para guardar el contenido de `student.new` en un archivo.

```
write.table(students.new, file = "student_new.txt", sep = "\t")
```

3. Lists

*Las listas son colecciones de objetos que pueden tener modos diferentes (e.g. numéricos, vectores, arrays..) # Ejemplo de cómo crear una lista. Ejecuta los comandos y describe que es lo que ocurre

```
# Crea una lista con los atributos que hemos definido y los valores que le hemos indicado
my_list <- list(name="Fred", wife="Mary", no.children=3, child.ages=c(4,7,9))

# Muestran los nombres de los atributos de la lista
attributes(my_list)

## $names
## [1] "name"      "wife"      "no.children" "child.ages"
names(my_list)
```

```
## [1] "name"          "wife"          "no.children" "child.ages"
# Muestra el contenido del segundo atributo de la lista
my_list[2]

## $wife
## [1] "Mary"
# Muestra el contenido segundo campo del segundo atributo de la lista
my_list[[2]]

## [1] "Mary"
# Muestra el contenido del atributo wife de nuestra lista
my_list$wife

## [1] "Mary"
# Muestra el segundo elemento contenido en el cuarto atributo de nuestra lista, child.ages.
my_list[[4]][2]

## [1] 7
#Imprime la longitud del cuarto atributo de la lista, child.ages.
length(my_list[[4]])

## [1] 3
#Sustituye el contenido del atributo wife de la lista por un array de valores del 1 al 12
my_list$wife <- 1:12

#Sustituye el contenido del atributo wife de la lista por Null
my_list$wife <- NULL

#Añade a nuestra lista un nuevo atributo con el nombre de los meses del año
my_list <- c(my_list, list(my_title2=month.name[1:12]))

#Deshace nuestra lista transformandola en un vector
unlist(my_list)

##          name no.children child.ages1 child.ages2 child.ages3 my_title21
##      "Fred"      "3"      "4"      "7"      "9"      "January"
## my_title22 my_title23 my_title24 my_title25 my_title26 my_title27
## "February" "March"   "April"   "May"    "June"   "July"
## my_title28 my_title29 my_title210 my_title211 my_title212
##  "August"  "September" "October" "November" "December"
#Crea un dataframe con el vector resultante de deshacer nuestra lista
data.frame(unlist(my_list))

##          unlist.my_list.
## name          Fred
## no.children          3
## child.ages1          4
## child.ages2          7
## child.ages3          9
## my_title21      January
## my_title22      February
## my_title23          March
## my_title24          April
```

```
## my_title25      May
## my_title26      June
## my_title27      July
## my_title28      August
## my_title29      September
## my_title210     October
## my_title211     November
## my_title212     December

#Crea una matriz con el vector resultante de deshacer nuestra lista
matrix(unlist(my_list))
```

```
##      [,1]
## [1,] "Fred"
## [2,] "3"
## [3,] "4"
## [4,] "7"
## [5,] "9"
## [6,] "January"
## [7,] "February"
## [8,] "March"
## [9,] "April"
## [10,] "May"
## [11,] "June"
## [12,] "July"
## [13,] "August"
## [14,] "September"
## [15,] "October"
## [16,] "November"
## [17,] "December"
```

4. table()

* La función table() cuenta el numero de elementos repetidos en un vector. Es la función más básica de clustering.

Cuenta el numero de entradas idénticas en la variable Sepal.Length del dataset iris.

```
table(iris$Sepal.Length)

##
## 4.3 4.4 4.5 4.6 4.7 4.8 4.9  5 5.1 5.2 5.3 5.4 5.5 5.6 5.7 5.8 5.9  6
##  1  3  1  4  2  5  6 10  9  4  1  6  7  6  8  7  3  6
## 6.1 6.2 6.3 6.4 6.5 6.6 6.7 6.8 6.9  7 7.1 7.2 7.3 7.4 7.6 7.7 7.9
##  6  4  9  7  5  2  8  3  4  1  1  3  1  1  1  4  1
```

5. Como ordenar datos, hacer selecciones con if(), calcular condicionales totales, transponer columnas y filas

Vamos a volver a utilizar el datasets mtcars.

-Ordena este data set de forma ascendente según su valor de hp. PISTA: with()

```
mtcars[with(mtcars, order(mtcars$hp)), ]
```

##	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
## Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
## Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
## Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
## AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
## Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
## Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
## Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4
## Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
## Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8

-Hazlo ahora de forma descendente

```
mtcars[with(mtcars, order(mtcars$hp, decreasing = TRUE)), ]
```

##	mpg	cyl	disp	hp	drat	wt	qsec	vs	am	gear	carb
## Maserati Bora	15.0	8	301.0	335	3.54	3.570	14.60	0	1	5	8
## Ford Pantera L	15.8	8	351.0	264	4.22	3.170	14.50	0	1	5	4
## Duster 360	14.3	8	360.0	245	3.21	3.570	15.84	0	0	3	4
## Camaro Z28	13.3	8	350.0	245	3.73	3.840	15.41	0	0	3	4

## Chrysler Imperial	14.7	8	440.0	230	3.23	5.345	17.42	0	0	3	4
## Lincoln Continental	10.4	8	460.0	215	3.00	5.424	17.82	0	0	3	4
## Cadillac Fleetwood	10.4	8	472.0	205	2.93	5.250	17.98	0	0	3	4
## Merc 450SE	16.4	8	275.8	180	3.07	4.070	17.40	0	0	3	3
## Merc 450SL	17.3	8	275.8	180	3.07	3.730	17.60	0	0	3	3
## Merc 450SLC	15.2	8	275.8	180	3.07	3.780	18.00	0	0	3	3
## Hornet Sportabout	18.7	8	360.0	175	3.15	3.440	17.02	0	0	3	2
## Pontiac Firebird	19.2	8	400.0	175	3.08	3.845	17.05	0	0	3	2
## Ferrari Dino	19.7	6	145.0	175	3.62	2.770	15.50	0	1	5	6
## Dodge Challenger	15.5	8	318.0	150	2.76	3.520	16.87	0	0	3	2
## AMC Javelin	15.2	8	304.0	150	3.15	3.435	17.30	0	0	3	2
## Merc 280	19.2	6	167.6	123	3.92	3.440	18.30	1	0	4	4
## Merc 280C	17.8	6	167.6	123	3.92	3.440	18.90	1	0	4	4
## Lotus Europa	30.4	4	95.1	113	3.77	1.513	16.90	1	1	5	2
## Mazda RX4	21.0	6	160.0	110	3.90	2.620	16.46	0	1	4	4
## Mazda RX4 Wag	21.0	6	160.0	110	3.90	2.875	17.02	0	1	4	4
## Hornet 4 Drive	21.4	6	258.0	110	3.08	3.215	19.44	1	0	3	1
## Volvo 142E	21.4	4	121.0	109	4.11	2.780	18.60	1	1	4	2
## Valiant	18.1	6	225.0	105	2.76	3.460	20.22	1	0	3	1
## Toyota Corona	21.5	4	120.1	97	3.70	2.465	20.01	1	0	3	1
## Merc 230	22.8	4	140.8	95	3.92	3.150	22.90	1	0	4	2
## Datsun 710	22.8	4	108.0	93	3.85	2.320	18.61	1	1	4	1
## Porsche 914-2	26.0	4	120.3	91	4.43	2.140	16.70	0	1	5	2
## Fiat 128	32.4	4	78.7	66	4.08	2.200	19.47	1	1	4	1
## Fiat X1-9	27.3	4	79.0	66	4.08	1.935	18.90	1	1	4	1
## Toyota Corolla	33.9	4	71.1	65	4.22	1.835	19.90	1	1	4	1
## Merc 240D	24.4	4	146.7	62	3.69	3.190	20.00	1	0	4	2
## Honda Civic	30.4	4	75.7	52	4.93	1.615	18.52	1	1	4	2

Calcula la media de la columna mpg

```
mean(mtcars$mpg)
```

```
## [1] 20.09062
```

Calcula la media de mpg para aquellos datos cuyo valor de hp sea menor que 150 y por separado para aquellos cuyo valor de hp sea mayor o igual a 150

```
#Media menor que 150
mean(mtcars[which(mtcars$hp < 150),]$mpg)
```

```
## [1] 24.22353
```

```
#Media mayor o igual que 150
mean(mtcars[which(mtcars$hp >= 150),]$mpg)
```

```
## [1] 15.40667
```

Busca los valores únicos de la columna cyl de mtcars. PISTA unique()

```
unique(mtcars$cyl)
```

```
## [1] 6 4 8
```

Obten los datos de mpg cyl disp hp para “Toyota Corolla”

```
mtcars["Toyota Corolla",c(1:4)]
```

```
##           mpg cyl disp hp
## Toyota Corolla 33.9   4 71.1 65
```

Crea una nueva variable mpgClass de tipo categórico cuyo valor es “Low” si el valor de mpg es menor que la media de la columna mpg y “High” si es mayor que la media de mpg. PISTA ifelse(). Combina ese comando con with() para añadir la nueva variable a mtcars.

```
mpg.new = data.frame(mtcars, mpgClass= with(mtcars, ifelse(mpg < mean(mpg), "Low", "High")))
```

¿qué pasa cuando ejecutas este comando?

Cuando ejecutas el comando with evalúa la expresión que pongamos sobre el data frame pudiendo modificar una copia del data frame original. En el ejercicio anterior, hemos usado with para evaluar la función ifelse sobre el data frame mtcars, obteniendo el resultado que hemos concatenado con el dataframe original para crear un nuevo data frame.