

Ejercicios 3 R

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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 subset de datos donde ambos Castores están activos.

```
beaver1$ID = 1
beaver2$ID = 2
beavers = rbind(beaver1,beaver2)
active_beavers = subset(beavers,activ == 1); active_beavers
```

```
##      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

```
nrow(USArrests)
```

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

- Numero 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 dataframe.

El primer elemento es el nombre de la fila, el segundo elemento es el nombre de las columnas.

```
dimnames(USArrests)
```

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

```
## [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"
```

```
##
```

```
## [[2]]
```

```
## [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 dataframe según el porcentaje de población en el área urbana. Para ello investiga `order()` y sus parámetros.

```
ord_usarrests = USArrests[order(USArrests[, "UrbanPop"], decreasing = T),]
```

```
ord_usarrests
```

```
##      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?

```
ord_usarrests2 = USArrests[order(USArrests$Murder,USArrests$UrbanPop,decreasing = T),]
ord_usarrests2
```

##	Murder	Assault	UrbanPop	Rape
## Georgia	17.4	211	60	25.8
## Mississippi	16.1	259	44	17.1
## Florida	15.4	335	80	31.9
## Louisiana	15.4	249	66	22.2
## South Carolina	14.4	279	48	22.5

```
## Tennessee      13.2      188      59 26.9
## Alabama         13.2      236      58 21.2
## North Carolina  13.0      337      45 16.1
## Texas           12.7      201      80 25.5
## Nevada          12.2      252      81 46.0
## Michigan        12.1      255      74 35.1
## New Mexico      11.4      285      70 32.1
## Maryland        11.3      300      67 27.8
## New York        11.1      254      86 26.1
## Illinois        10.4      249      83 24.0
## Alaska          10.0      263      48 44.5
## Kentucky        9.7       109      52 16.3
## California      9.0       276      91 40.6
## Missouri        9.0       178      70 28.2
## Arkansas        8.8       190      50 19.5
## Virginia        8.5       156      63 20.7
## Arizona         8.1       294      80 31.0
## Colorado        7.9       204      78 38.7
## New Jersey      7.4       159      89 18.8
## Ohio            7.3       120      75 21.4
## Indiana         7.2       113      65 21.0
## Wyoming         6.8       161      60 15.6
## Oklahoma        6.6       151      68 20.0
## Pennsylvania    6.3       106      72 14.9
## Kansas          6.0       115      66 18.0
## Montana         6.0       109      53 16.4
## Delaware        5.9       238      72 15.8
## West Virginia   5.7        81      39 9.3
## Hawaii          5.3        46      83 20.2
## Oregon          4.9       159      67 29.3
## Massachusetts   4.4       149      85 16.3
## Nebraska        4.3       102      62 16.5
## Washington      4.0       145      73 26.2
## South Dakota    3.8        86      45 12.8
## Rhode Island    3.4       174      87 8.3
## Connecticut     3.3       110      77 11.1
## Utah            3.2       120      80 22.9
## Minnesota       2.7        72      66 14.9
## Wisconsin       2.6        53      66 10.8
## Idaho           2.6       120      54 14.2
## Iowa            2.2        56      57 11.3
## Vermont         2.2        48      32 11.2
## New Hampshire   2.1        57      56 9.5
## Maine           2.1        83      51 7.8
## North Dakota    0.8        45      44 7.3
```

- 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 todas las filas para las dos primeras columnas.

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

```
##           Murder Assault UrbanPop Rape
## Alabama    13.2     236         58 21.2
## Alaska     10.0     263         48 44.5
```

- Muestra todas las filas de la 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
```

- ¿Qué estado tiene la menor tasa de asesinatos? ¿Qué línea contiene esa información? Obtén esa información.

```
USArrests[which.min(USArrests[, "Murder"]),]
```

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

```
which.min(USArrests[, "Murder"])
```

```
## [1] 34
```

- ¿Qué estados tienen una tasa inferior al 4%?, obtén esa información.

```
USArrests[which(USArrests[, "Murder"] < 4.0),]
```

```
##           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
```

- ¿Qué estados están en el cuartil superior(75) en lo que a población en zonas urbanas se refiere?

```
USArrests[USArrests$UrbanPop >= 75,]
```

```
##           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
## Connecticut   3.3      110      77 11.1
## 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
## Ohio           7.3      120      75 21.4
## 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.

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

```
str(CO2)
```

```
## Classes 'nfnGroupedData', 'nfGroupedData', 'groupedData' and 'data.frame':  84 obs. of  5 variables
## $ Plant      : Ord.factor w/ 12 levels "Qn1"<"Qn2"<"Qn3"<...: 1 1 1 1 1 1 1 2 2 2 ...
## $ Type       : Factor w/ 2 levels "Quebec","Mississippi": 1 1 1 1 1 1 1 1 1 1 ...
## $ Treatment: Factor w/ 2 levels "nonchilled","chilled": 1 1 1 1 1 1 1 1 1 1 ...
## $ conc       : num  95 175 250 350 500 675 1000 95 175 250 ...
## $ uptake     : num  16 30.4 34.8 37.2 35.3 39.2 39.7 13.6 27.3 37.1 ...
## - attr(*, "formula")=Class 'formula' language uptake ~ conc | Plant
## .. ..- attr(*, ".Environment")=<environment: R_EmptyEnv>
## - attr(*, "outer")=Class 'formula' language ~Treatment * Type
## .. ..- attr(*, ".Environment")=<environment: R_EmptyEnv>
## - attr(*, "labels")=List of 2
## ..$ x: chr "Ambient carbon dioxide concentration"
## ..$ y: chr "CO2 uptake rate"
## - attr(*, "units")=List of 2
## ..$ x: chr "(uL/L)"
## ..$ y: chr "(umol/m^2 s)"
```

```
CO2[order(CO2$Plant,decreasing = F),]
```

```
##   Plant      Type Treatment conc uptake
## 1   Qn1   Quebec nonchilled   95   16.0
## 2   Qn1   Quebec nonchilled  175   30.4
## 3   Qn1   Quebec nonchilled  250   34.8
## 4   Qn1   Quebec nonchilled  350   37.2
## 5   Qn1   Quebec nonchilled  500   35.3
## 6   Qn1   Quebec nonchilled  675   39.2
## 7   Qn1   Quebec nonchilled 1000   39.7
## 8   Qn2   Quebec nonchilled   95   13.6
## 9   Qn2   Quebec nonchilled  175   27.3
## 10  Qn2   Quebec nonchilled  250   37.1
## 11  Qn2   Quebec nonchilled  350   41.8
```


## 12	Qn2	Quebec nonchilled	500	40.6
## 13	Qn2	Quebec nonchilled	675	41.4
## 14	Qn2	Quebec nonchilled	1000	44.3
## 15	Qn3	Quebec nonchilled	95	16.2
## 16	Qn3	Quebec nonchilled	175	32.4
## 17	Qn3	Quebec nonchilled	250	40.3
## 18	Qn3	Quebec nonchilled	350	42.1
## 19	Qn3	Quebec nonchilled	500	42.9
## 20	Qn3	Quebec nonchilled	675	43.9
## 21	Qn3	Quebec nonchilled	1000	45.5
## 22	Qc1	Quebec chilled	95	14.2
## 23	Qc1	Quebec chilled	175	24.1
## 24	Qc1	Quebec chilled	250	30.3
## 25	Qc1	Quebec chilled	350	34.6
## 26	Qc1	Quebec chilled	500	32.5
## 27	Qc1	Quebec chilled	675	35.4
## 28	Qc1	Quebec chilled	1000	38.7
## 36	Qc3	Quebec chilled	95	15.1
## 37	Qc3	Quebec chilled	175	21.0
## 38	Qc3	Quebec chilled	250	38.1
## 39	Qc3	Quebec chilled	350	34.0
## 40	Qc3	Quebec chilled	500	38.9
## 41	Qc3	Quebec chilled	675	39.6
## 42	Qc3	Quebec chilled	1000	41.4
## 29	Qc2	Quebec chilled	95	9.3
## 30	Qc2	Quebec chilled	175	27.3
## 31	Qc2	Quebec chilled	250	35.0
## 32	Qc2	Quebec chilled	350	38.8
## 33	Qc2	Quebec chilled	500	38.6
## 34	Qc2	Quebec chilled	675	37.5
## 35	Qc2	Quebec chilled	1000	42.4
## 57	Mn3	Mississippi nonchilled	95	11.3
## 58	Mn3	Mississippi nonchilled	175	19.4
## 59	Mn3	Mississippi nonchilled	250	25.8
## 60	Mn3	Mississippi nonchilled	350	27.9
## 61	Mn3	Mississippi nonchilled	500	28.5
## 62	Mn3	Mississippi nonchilled	675	28.1
## 63	Mn3	Mississippi nonchilled	1000	27.8
## 50	Mn2	Mississippi nonchilled	95	12.0
## 51	Mn2	Mississippi nonchilled	175	22.0
## 52	Mn2	Mississippi nonchilled	250	30.6
## 53	Mn2	Mississippi nonchilled	350	31.8
## 54	Mn2	Mississippi nonchilled	500	32.4
## 55	Mn2	Mississippi nonchilled	675	31.1
## 56	Mn2	Mississippi nonchilled	1000	31.5
## 43	Mn1	Mississippi nonchilled	95	10.6
## 44	Mn1	Mississippi nonchilled	175	19.2
## 45	Mn1	Mississippi nonchilled	250	26.2
## 46	Mn1	Mississippi nonchilled	350	30.0
## 47	Mn1	Mississippi nonchilled	500	30.9
## 48	Mn1	Mississippi nonchilled	675	32.4
## 49	Mn1	Mississippi nonchilled	1000	35.5
## 71	Mc2	Mississippi chilled	95	7.7
## 72	Mc2	Mississippi chilled	175	11.4

```
## 73 Mc2 Mississippi chilled 250 12.3
## 74 Mc2 Mississippi chilled 350 13.0
## 75 Mc2 Mississippi chilled 500 12.5
## 76 Mc2 Mississippi chilled 675 13.7
## 77 Mc2 Mississippi chilled 1000 14.4
## 78 Mc3 Mississippi chilled 95 10.6
## 79 Mc3 Mississippi chilled 175 18.0
## 80 Mc3 Mississippi chilled 250 17.9
## 81 Mc3 Mississippi chilled 350 17.9
## 82 Mc3 Mississippi chilled 500 17.9
## 83 Mc3 Mississippi chilled 675 18.9
## 84 Mc3 Mississippi chilled 1000 19.9
## 64 Mc1 Mississippi chilled 95 10.5
## 65 Mc1 Mississippi chilled 175 14.9
## 66 Mc1 Mississippi chilled 250 18.1
## 67 Mc1 Mississippi chilled 350 18.9
## 68 Mc1 Mississippi chilled 500 19.5
## 69 Mc1 Mississippi chilled 675 22.2
## 70 Mc1 Mississippi chilled 1000 21.9
```

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

```
# Ordenamos según incremento de uptake
new_co2 = CO2
new_co2 = new_co2[order(new_co2$uptake),]
new_co2
```

```
## Plant Type Treatment conc uptake
## 71 Mc2 Mississippi chilled 95 7.7
## 29 Qc2 Quebec chilled 95 9.3
## 64 Mc1 Mississippi chilled 95 10.5
## 43 Mn1 Mississippi nonchilled 95 10.6
## 78 Mc3 Mississippi chilled 95 10.6
## 57 Mn3 Mississippi nonchilled 95 11.3
## 72 Mc2 Mississippi chilled 175 11.4
## 50 Mn2 Mississippi nonchilled 95 12.0
## 73 Mc2 Mississippi chilled 250 12.3
## 75 Mc2 Mississippi chilled 500 12.5
## 74 Mc2 Mississippi chilled 350 13.0
## 8 Qn2 Quebec nonchilled 95 13.6
## 76 Mc2 Mississippi chilled 675 13.7
## 22 Qc1 Quebec chilled 95 14.2
## 77 Mc2 Mississippi chilled 1000 14.4
## 65 Mc1 Mississippi chilled 175 14.9
## 36 Qc3 Quebec chilled 95 15.1
## 1 Qn1 Quebec nonchilled 95 16.0
## 15 Qn3 Quebec nonchilled 95 16.2
## 80 Mc3 Mississippi chilled 250 17.9
## 81 Mc3 Mississippi chilled 350 17.9
## 82 Mc3 Mississippi chilled 500 17.9
## 79 Mc3 Mississippi chilled 175 18.0
## 66 Mc1 Mississippi chilled 250 18.1
## 67 Mc1 Mississippi chilled 350 18.9
## 83 Mc3 Mississippi chilled 675 18.9
```

## 44	Mn1	Mississippi	nonchilled	175	19.2
## 58	Mn3	Mississippi	nonchilled	175	19.4
## 68	Mc1	Mississippi	chilled	500	19.5
## 84	Mc3	Mississippi	chilled	1000	19.9
## 37	Qc3	Quebec	chilled	175	21.0
## 70	Mc1	Mississippi	chilled	1000	21.9
## 51	Mn2	Mississippi	nonchilled	175	22.0
## 69	Mc1	Mississippi	chilled	675	22.2
## 23	Qc1	Quebec	chilled	175	24.1
## 59	Mn3	Mississippi	nonchilled	250	25.8
## 45	Mn1	Mississippi	nonchilled	250	26.2
## 9	Qn2	Quebec	nonchilled	175	27.3
## 30	Qc2	Quebec	chilled	175	27.3
## 63	Mn3	Mississippi	nonchilled	1000	27.8
## 60	Mn3	Mississippi	nonchilled	350	27.9
## 62	Mn3	Mississippi	nonchilled	675	28.1
## 61	Mn3	Mississippi	nonchilled	500	28.5
## 46	Mn1	Mississippi	nonchilled	350	30.0
## 24	Qc1	Quebec	chilled	250	30.3
## 2	Qn1	Quebec	nonchilled	175	30.4
## 52	Mn2	Mississippi	nonchilled	250	30.6
## 47	Mn1	Mississippi	nonchilled	500	30.9
## 55	Mn2	Mississippi	nonchilled	675	31.1
## 56	Mn2	Mississippi	nonchilled	1000	31.5
## 53	Mn2	Mississippi	nonchilled	350	31.8
## 16	Qn3	Quebec	nonchilled	175	32.4
## 48	Mn1	Mississippi	nonchilled	675	32.4
## 54	Mn2	Mississippi	nonchilled	500	32.4
## 26	Qc1	Quebec	chilled	500	32.5
## 39	Qc3	Quebec	chilled	350	34.0
## 25	Qc1	Quebec	chilled	350	34.6
## 3	Qn1	Quebec	nonchilled	250	34.8
## 31	Qc2	Quebec	chilled	250	35.0
## 5	Qn1	Quebec	nonchilled	500	35.3
## 27	Qc1	Quebec	chilled	675	35.4
## 49	Mn1	Mississippi	nonchilled	1000	35.5
## 10	Qn2	Quebec	nonchilled	250	37.1
## 4	Qn1	Quebec	nonchilled	350	37.2
## 34	Qc2	Quebec	chilled	675	37.5
## 38	Qc3	Quebec	chilled	250	38.1
## 33	Qc2	Quebec	chilled	500	38.6
## 28	Qc1	Quebec	chilled	1000	38.7
## 32	Qc2	Quebec	chilled	350	38.8
## 40	Qc3	Quebec	chilled	500	38.9
## 6	Qn1	Quebec	nonchilled	675	39.2
## 41	Qc3	Quebec	chilled	675	39.6
## 7	Qn1	Quebec	nonchilled	1000	39.7
## 17	Qn3	Quebec	nonchilled	250	40.3
## 12	Qn2	Quebec	nonchilled	500	40.6
## 13	Qn2	Quebec	nonchilled	675	41.4
## 42	Qc3	Quebec	chilled	1000	41.4
## 11	Qn2	Quebec	nonchilled	350	41.8
## 18	Qn3	Quebec	nonchilled	350	42.1
## 35	Qc2	Quebec	chilled	1000	42.4

```
## 19  Qn3      Quebec nonchilled  500  42.9
## 20  Qn3      Quebec nonchilled  675  43.9
## 14  Qn2      Quebec nonchilled 1000  44.3
## 21  Qn3      Quebec nonchilled 1000  45.5
```

```
# Ordenamos alfabéticamente.
```

```
nuevo_factor = factor(new_co2$Plant, levels = levels(new_co2$Plant)[order(levels(new_co2$Plant))], ordered = TRUE)
new_co2 = new_co2[order(nuevo_factor),]
new_co2
```

##	Plant	Type	Treatment	conc	uptake
## 64	Mc1	Mississippi	chilled	95	10.5
## 65	Mc1	Mississippi	chilled	175	14.9
## 66	Mc1	Mississippi	chilled	250	18.1
## 67	Mc1	Mississippi	chilled	350	18.9
## 68	Mc1	Mississippi	chilled	500	19.5
## 70	Mc1	Mississippi	chilled	1000	21.9
## 69	Mc1	Mississippi	chilled	675	22.2
## 71	Mc2	Mississippi	chilled	95	7.7
## 72	Mc2	Mississippi	chilled	175	11.4
## 73	Mc2	Mississippi	chilled	250	12.3
## 75	Mc2	Mississippi	chilled	500	12.5
## 74	Mc2	Mississippi	chilled	350	13.0
## 76	Mc2	Mississippi	chilled	675	13.7
## 77	Mc2	Mississippi	chilled	1000	14.4
## 78	Mc3	Mississippi	chilled	95	10.6
## 80	Mc3	Mississippi	chilled	250	17.9
## 81	Mc3	Mississippi	chilled	350	17.9
## 82	Mc3	Mississippi	chilled	500	17.9
## 79	Mc3	Mississippi	chilled	175	18.0
## 83	Mc3	Mississippi	chilled	675	18.9
## 84	Mc3	Mississippi	chilled	1000	19.9
## 43	Mn1	Mississippi	nonchilled	95	10.6
## 44	Mn1	Mississippi	nonchilled	175	19.2
## 45	Mn1	Mississippi	nonchilled	250	26.2
## 46	Mn1	Mississippi	nonchilled	350	30.0
## 47	Mn1	Mississippi	nonchilled	500	30.9
## 48	Mn1	Mississippi	nonchilled	675	32.4
## 49	Mn1	Mississippi	nonchilled	1000	35.5
## 50	Mn2	Mississippi	nonchilled	95	12.0
## 51	Mn2	Mississippi	nonchilled	175	22.0
## 52	Mn2	Mississippi	nonchilled	250	30.6
## 55	Mn2	Mississippi	nonchilled	675	31.1
## 56	Mn2	Mississippi	nonchilled	1000	31.5
## 53	Mn2	Mississippi	nonchilled	350	31.8
## 54	Mn2	Mississippi	nonchilled	500	32.4
## 57	Mn3	Mississippi	nonchilled	95	11.3
## 58	Mn3	Mississippi	nonchilled	175	19.4
## 59	Mn3	Mississippi	nonchilled	250	25.8
## 63	Mn3	Mississippi	nonchilled	1000	27.8
## 60	Mn3	Mississippi	nonchilled	350	27.9
## 62	Mn3	Mississippi	nonchilled	675	28.1
## 61	Mn3	Mississippi	nonchilled	500	28.5
## 22	Qc1	Quebec	chilled	95	14.2
## 23	Qc1	Quebec	chilled	175	24.1

##	24	Qc1	Quebec	chilled	250	30.3
##	26	Qc1	Quebec	chilled	500	32.5
##	25	Qc1	Quebec	chilled	350	34.6
##	27	Qc1	Quebec	chilled	675	35.4
##	28	Qc1	Quebec	chilled	1000	38.7
##	29	Qc2	Quebec	chilled	95	9.3
##	30	Qc2	Quebec	chilled	175	27.3
##	31	Qc2	Quebec	chilled	250	35.0
##	34	Qc2	Quebec	chilled	675	37.5
##	33	Qc2	Quebec	chilled	500	38.6
##	32	Qc2	Quebec	chilled	350	38.8
##	35	Qc2	Quebec	chilled	1000	42.4
##	36	Qc3	Quebec	chilled	95	15.1
##	37	Qc3	Quebec	chilled	175	21.0
##	39	Qc3	Quebec	chilled	350	34.0
##	38	Qc3	Quebec	chilled	250	38.1
##	40	Qc3	Quebec	chilled	500	38.9
##	41	Qc3	Quebec	chilled	675	39.6
##	42	Qc3	Quebec	chilled	1000	41.4
##	1	Qn1	Quebec	nonchilled	95	16.0
##	2	Qn1	Quebec	nonchilled	175	30.4
##	3	Qn1	Quebec	nonchilled	250	34.8
##	5	Qn1	Quebec	nonchilled	500	35.3
##	4	Qn1	Quebec	nonchilled	350	37.2
##	6	Qn1	Quebec	nonchilled	675	39.2
##	7	Qn1	Quebec	nonchilled	1000	39.7
##	8	Qn2	Quebec	nonchilled	95	13.6
##	9	Qn2	Quebec	nonchilled	175	27.3
##	10	Qn2	Quebec	nonchilled	250	37.1
##	12	Qn2	Quebec	nonchilled	500	40.6
##	13	Qn2	Quebec	nonchilled	675	41.4
##	11	Qn2	Quebec	nonchilled	350	41.8
##	14	Qn2	Quebec	nonchilled	1000	44.3
##	15	Qn3	Quebec	nonchilled	95	16.2
##	16	Qn3	Quebec	nonchilled	175	32.4
##	17	Qn3	Quebec	nonchilled	250	40.3
##	18	Qn3	Quebec	nonchilled	350	42.1
##	19	Qn3	Quebec	nonchilled	500	42.9
##	20	Qn3	Quebec	nonchilled	675	43.9
##	21	Qn3	Quebec	nonchilled	1000	45.5

- Ordena de nuevo los datos en función del incremento de la variable uptake y el orden alfabético reverso de la planta (en ese orden)

```
# Ordenamos según incremento de uptake
new_co2 = CO2
new_co2 = new_co2[order(new_co2$uptake),]
new_co2
```

##	Plant	Type	Treatment	conc	uptake	
##	71	Mc2	Mississippi	chilled	95	7.7
##	29	Qc2	Quebec	chilled	95	9.3
##	64	Mc1	Mississippi	chilled	95	10.5
##	43	Mn1	Mississippi	nonchilled	95	10.6
##	78	Mc3	Mississippi	chilled	95	10.6

## 57	Mn3	Mississippi	nonchilled	95	11.3
## 72	Mc2	Mississippi	chilled	175	11.4
## 50	Mn2	Mississippi	nonchilled	95	12.0
## 73	Mc2	Mississippi	chilled	250	12.3
## 75	Mc2	Mississippi	chilled	500	12.5
## 74	Mc2	Mississippi	chilled	350	13.0
## 8	Qn2	Quebec	nonchilled	95	13.6
## 76	Mc2	Mississippi	chilled	675	13.7
## 22	Qc1	Quebec	chilled	95	14.2
## 77	Mc2	Mississippi	chilled	1000	14.4
## 65	Mc1	Mississippi	chilled	175	14.9
## 36	Qc3	Quebec	chilled	95	15.1
## 1	Qn1	Quebec	nonchilled	95	16.0
## 15	Qn3	Quebec	nonchilled	95	16.2
## 80	Mc3	Mississippi	chilled	250	17.9
## 81	Mc3	Mississippi	chilled	350	17.9
## 82	Mc3	Mississippi	chilled	500	17.9
## 79	Mc3	Mississippi	chilled	175	18.0
## 66	Mc1	Mississippi	chilled	250	18.1
## 67	Mc1	Mississippi	chilled	350	18.9
## 83	Mc3	Mississippi	chilled	675	18.9
## 44	Mn1	Mississippi	nonchilled	175	19.2
## 58	Mn3	Mississippi	nonchilled	175	19.4
## 68	Mc1	Mississippi	chilled	500	19.5
## 84	Mc3	Mississippi	chilled	1000	19.9
## 37	Qc3	Quebec	chilled	175	21.0
## 70	Mc1	Mississippi	chilled	1000	21.9
## 51	Mn2	Mississippi	nonchilled	175	22.0
## 69	Mc1	Mississippi	chilled	675	22.2
## 23	Qc1	Quebec	chilled	175	24.1
## 59	Mn3	Mississippi	nonchilled	250	25.8
## 45	Mn1	Mississippi	nonchilled	250	26.2
## 9	Qn2	Quebec	nonchilled	175	27.3
## 30	Qc2	Quebec	chilled	175	27.3
## 63	Mn3	Mississippi	nonchilled	1000	27.8
## 60	Mn3	Mississippi	nonchilled	350	27.9
## 62	Mn3	Mississippi	nonchilled	675	28.1
## 61	Mn3	Mississippi	nonchilled	500	28.5
## 46	Mn1	Mississippi	nonchilled	350	30.0
## 24	Qc1	Quebec	chilled	250	30.3
## 2	Qn1	Quebec	nonchilled	175	30.4
## 52	Mn2	Mississippi	nonchilled	250	30.6
## 47	Mn1	Mississippi	nonchilled	500	30.9
## 55	Mn2	Mississippi	nonchilled	675	31.1
## 56	Mn2	Mississippi	nonchilled	1000	31.5
## 53	Mn2	Mississippi	nonchilled	350	31.8
## 16	Qn3	Quebec	nonchilled	175	32.4
## 48	Mn1	Mississippi	nonchilled	675	32.4
## 54	Mn2	Mississippi	nonchilled	500	32.4
## 26	Qc1	Quebec	chilled	500	32.5
## 39	Qc3	Quebec	chilled	350	34.0
## 25	Qc1	Quebec	chilled	350	34.6
## 3	Qn1	Quebec	nonchilled	250	34.8
## 31	Qc2	Quebec	chilled	250	35.0

```
## 5    Qn1      Quebec nonchilled 500 35.3
## 27   Qc1      Quebec   chilled 675 35.4
## 49   Mn1 Mississippi nonchilled 1000 35.5
## 10   Qn2      Quebec nonchilled 250 37.1
## 4    Qn1      Quebec nonchilled 350 37.2
## 34   Qc2      Quebec   chilled 675 37.5
## 38   Qc3      Quebec   chilled 250 38.1
## 33   Qc2      Quebec   chilled 500 38.6
## 28   Qc1      Quebec   chilled 1000 38.7
## 32   Qc2      Quebec   chilled 350 38.8
## 40   Qc3      Quebec   chilled 500 38.9
## 6    Qn1      Quebec nonchilled 675 39.2
## 41   Qc3      Quebec   chilled 675 39.6
## 7    Qn1      Quebec nonchilled 1000 39.7
## 17   Qn3      Quebec nonchilled 250 40.3
## 12   Qn2      Quebec nonchilled 500 40.6
## 13   Qn2      Quebec nonchilled 675 41.4
## 42   Qc3      Quebec   chilled 1000 41.4
## 11   Qn2      Quebec nonchilled 350 41.8
## 18   Qn3      Quebec nonchilled 350 42.1
## 35   Qc2      Quebec   chilled 1000 42.4
## 19   Qn3      Quebec nonchilled 500 42.9
## 20   Qn3      Quebec nonchilled 675 43.9
## 14   Qn2      Quebec nonchilled 1000 44.3
## 21   Qn3      Quebec nonchilled 1000 45.5
```

```
# Ordenamos en orden alfabético reverso
```

```
nuevo_factor = factor(new_co2$Plant, levels = levels(new_co2$Plant)[order(levels(new_co2$Plant), decreasing = TRUE)])
new_co2 = new_co2[order(nuevo_factor),]
new_co2
```

```
##      Plant      Type Treatment conc uptake
## 15   Qn3      Quebec nonchilled   95  16.2
## 16   Qn3      Quebec nonchilled  175  32.4
## 17   Qn3      Quebec nonchilled  250  40.3
## 18   Qn3      Quebec nonchilled  350  42.1
## 19   Qn3      Quebec nonchilled  500  42.9
## 20   Qn3      Quebec nonchilled  675  43.9
## 21   Qn3      Quebec nonchilled 1000  45.5
## 8    Qn2      Quebec nonchilled   95  13.6
## 9    Qn2      Quebec nonchilled  175  27.3
## 10   Qn2      Quebec nonchilled  250  37.1
## 12   Qn2      Quebec nonchilled  500  40.6
## 13   Qn2      Quebec nonchilled  675  41.4
## 11   Qn2      Quebec nonchilled  350  41.8
## 14   Qn2      Quebec nonchilled 1000  44.3
## 1    Qn1      Quebec nonchilled   95  16.0
## 2    Qn1      Quebec nonchilled  175  30.4
## 3    Qn1      Quebec nonchilled  250  34.8
## 5    Qn1      Quebec nonchilled  500  35.3
## 4    Qn1      Quebec nonchilled  350  37.2
## 6    Qn1      Quebec nonchilled  675  39.2
## 7    Qn1      Quebec nonchilled 1000  39.7
## 36   Qc3      Quebec   chilled   95  15.1
## 37   Qc3      Quebec   chilled  175  21.0
```

## 39	Qc3	Quebec	chilled	350	34.0
## 38	Qc3	Quebec	chilled	250	38.1
## 40	Qc3	Quebec	chilled	500	38.9
## 41	Qc3	Quebec	chilled	675	39.6
## 42	Qc3	Quebec	chilled	1000	41.4
## 29	Qc2	Quebec	chilled	95	9.3
## 30	Qc2	Quebec	chilled	175	27.3
## 31	Qc2	Quebec	chilled	250	35.0
## 34	Qc2	Quebec	chilled	675	37.5
## 33	Qc2	Quebec	chilled	500	38.6
## 32	Qc2	Quebec	chilled	350	38.8
## 35	Qc2	Quebec	chilled	1000	42.4
## 22	Qc1	Quebec	chilled	95	14.2
## 23	Qc1	Quebec	chilled	175	24.1
## 24	Qc1	Quebec	chilled	250	30.3
## 26	Qc1	Quebec	chilled	500	32.5
## 25	Qc1	Quebec	chilled	350	34.6
## 27	Qc1	Quebec	chilled	675	35.4
## 28	Qc1	Quebec	chilled	1000	38.7
## 57	Mn3	Mississippi	nonchilled	95	11.3
## 58	Mn3	Mississippi	nonchilled	175	19.4
## 59	Mn3	Mississippi	nonchilled	250	25.8
## 63	Mn3	Mississippi	nonchilled	1000	27.8
## 60	Mn3	Mississippi	nonchilled	350	27.9
## 62	Mn3	Mississippi	nonchilled	675	28.1
## 61	Mn3	Mississippi	nonchilled	500	28.5
## 50	Mn2	Mississippi	nonchilled	95	12.0
## 51	Mn2	Mississippi	nonchilled	175	22.0
## 52	Mn2	Mississippi	nonchilled	250	30.6
## 55	Mn2	Mississippi	nonchilled	675	31.1
## 56	Mn2	Mississippi	nonchilled	1000	31.5
## 53	Mn2	Mississippi	nonchilled	350	31.8
## 54	Mn2	Mississippi	nonchilled	500	32.4
## 43	Mn1	Mississippi	nonchilled	95	10.6
## 44	Mn1	Mississippi	nonchilled	175	19.2
## 45	Mn1	Mississippi	nonchilled	250	26.2
## 46	Mn1	Mississippi	nonchilled	350	30.0
## 47	Mn1	Mississippi	nonchilled	500	30.9
## 48	Mn1	Mississippi	nonchilled	675	32.4
## 49	Mn1	Mississippi	nonchilled	1000	35.5
## 78	Mc3	Mississippi	chilled	95	10.6
## 80	Mc3	Mississippi	chilled	250	17.9
## 81	Mc3	Mississippi	chilled	350	17.9
## 82	Mc3	Mississippi	chilled	500	17.9
## 79	Mc3	Mississippi	chilled	175	18.0
## 83	Mc3	Mississippi	chilled	675	18.9
## 84	Mc3	Mississippi	chilled	1000	19.9
## 71	Mc2	Mississippi	chilled	95	7.7
## 72	Mc2	Mississippi	chilled	175	11.4
## 73	Mc2	Mississippi	chilled	250	12.3
## 75	Mc2	Mississippi	chilled	500	12.5
## 74	Mc2	Mississippi	chilled	350	13.0
## 76	Mc2	Mississippi	chilled	675	13.7
## 77	Mc2	Mississippi	chilled	1000	14.4


```
## 64 Mc1 Mississippi chilled 95 10.5
## 65 Mc1 Mississippi chilled 175 14.9
## 66 Mc1 Mississippi chilled 250 18.1
## 67 Mc1 Mississippi chilled 350 18.9
## 68 Mc1 Mississippi chilled 500 19.5
## 70 Mc1 Mississippi chilled 1000 21.9
## 69 Mc1 Mississippi chilled 675 22.2
```

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"
```

```
statex77 = as.data.frame(state.x77)
```

```
str(statex77)
```

```
## 'data.frame': 50 obs. of 8 variables:
## $ Population: num 3615 365 2212 2110 21198 ...
## $ Income : num 3624 6315 4530 3378 5114 ...
## $ Illiteracy: num 2.1 1.5 1.8 1.9 1.1 0.7 1.1 0.9 1.3 2 ...
## $ Life Exp : num 69 69.3 70.5 70.7 71.7 ...
## $ Murder : num 15.1 11.3 7.8 10.1 10.3 6.8 3.1 6.2 10.7 13.9 ...
## $ HS Grad : num 41.3 66.7 58.1 39.9 62.6 63.9 56 54.6 52.6 40.6 ...
## $ Frost : num 20 152 15 65 20 166 139 103 11 60 ...
## $ Area : num 50708 566432 113417 51945 156361 ...
```

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

```
help("subset")
```

```
subset(statex77,statex77$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.

```
statex77[which.max(statex77$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.

```
new_names = unlist(strsplit(colnames(df2), "state."))
new_names = new_names[new_names != ""]; new_names
```

```
## [1] "abb"      "area"      "division" "region"
```

```
colnames(df2) = new_names
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

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

```

statex77 = cbind(statex77,df2)
drops = c("Life Exp","HS Grad","Frost","abb","area")
statex77 = statex77[,!colnames(statex77) %in% drops]
statex77

```

##	Population	Income	Illiteracy	Murder	Area
## Alabama	3615	3624	2.1	15.1	50708
## Alaska	365	6315	1.5	11.3	566432
## Arizona	2212	4530	1.8	7.8	113417
## Arkansas	2110	3378	1.9	10.1	51945
## California	21198	5114	1.1	10.3	156361
## Colorado	2541	4884	0.7	6.8	103766
## Connecticut	3100	5348	1.1	3.1	4862
## Delaware	579	4809	0.9	6.2	1982
## Florida	8277	4815	1.3	10.7	54090
## Georgia	4931	4091	2.0	13.9	58073
## Hawaii	868	4963	1.9	6.2	6425

## Idaho	813	4119	0.6	5.3	82677
## Illinois	11197	5107	0.9	10.3	55748
## Indiana	5313	4458	0.7	7.1	36097
## Iowa	2861	4628	0.5	2.3	55941
## Kansas	2280	4669	0.6	4.5	81787
## Kentucky	3387	3712	1.6	10.6	39650
## Louisiana	3806	3545	2.8	13.2	44930
## Maine	1058	3694	0.7	2.7	30920
## Maryland	4122	5299	0.9	8.5	9891
## Massachusetts	5814	4755	1.1	3.3	7826
## Michigan	9111	4751	0.9	11.1	56817
## Minnesota	3921	4675	0.6	2.3	79289
## Mississippi	2341	3098	2.4	12.5	47296
## Missouri	4767	4254	0.8	9.3	68995
## Montana	746	4347	0.6	5.0	145587
## Nebraska	1544	4508	0.6	2.9	76483
## Nevada	590	5149	0.5	11.5	109889
## New Hampshire	812	4281	0.7	3.3	9027
## New Jersey	7333	5237	1.1	5.2	7521
## New Mexico	1144	3601	2.2	9.7	121412
## New York	18076	4903	1.4	10.9	47831
## North Carolina	5441	3875	1.8	11.1	48798
## North Dakota	637	5087	0.8	1.4	69273
## Ohio	10735	4561	0.8	7.4	40975
## Oklahoma	2715	3983	1.1	6.4	68782
## Oregon	2284	4660	0.6	4.2	96184
## Pennsylvania	11860	4449	1.0	6.1	44966
## Rhode Island	931	4558	1.3	2.4	1049
## South Carolina	2816	3635	2.3	11.6	30225
## South Dakota	681	4167	0.5	1.7	75955
## Tennessee	4173	3821	1.7	11.0	41328
## Texas	12237	4188	2.2	12.2	262134
## Utah	1203	4022	0.6	4.5	82096
## Vermont	472	3907	0.6	5.5	9267
## Virginia	4981	4701	1.4	9.5	39780
## Washington	3559	4864	0.6	4.3	66570
## West Virginia	1799	3617	1.4	6.7	24070
## Wisconsin	4589	4468	0.7	3.0	54464
## Wyoming	376	4566	0.6	6.9	97203
##	division		region		
## Alabama	East	South Central	South		
## Alaska		Pacific	West		
## Arizona		Mountain	West		
## Arkansas	West	South Central	South		
## California		Pacific	West		
## Colorado		Mountain	West		
## Connecticut		New England	Northeast		
## Delaware		South Atlantic	South		
## Florida		South Atlantic	South		
## Georgia		South Atlantic	South		
## Hawaii		Pacific	West		
## Idaho		Mountain	West		
## Illinois	East	North Central	North Central		
## Indiana	East	North Central	North Central		

```
## Iowa          West North Central North Central
## Kansas        West North Central North Central
## Kentucky      East South Central      South
## Louisiana     West South Central      South
## Maine         New England      Northeast
## Maryland      South Atlantic      South
## Massachusetts New England      Northeast
## Michigan      East North Central North Central
## Minnesota     West North Central North Central
## Mississippi   East South Central      South
## Missouri      West North Central North Central
## Montana       Mountain      West
## Nebraska      West North Central North Central
## Nevada        Mountain      West
## New Hampshire New England      Northeast
## New Jersey    Middle Atlantic      Northeast
## New Mexico    Mountain      West
## New York      Middle Atlantic      Northeast
## North Carolina South Atlantic      South
## North Dakota  West North Central North Central
## Ohio          East North Central North Central
## Oklahoma      West South Central      South
## Oregon        Pacific      West
## Pennsylvania  Middle Atlantic      Northeast
## Rhode Island  New England      Northeast
## South Carolina South Atlantic      South
## South Dakota  West North Central North Central
## Tennessee     East South Central      South
## Texas         West South Central      South
## Utah          Mountain      West
## Vermont       New England      Northeast
## Virginia      South Atlantic      South
## Washington    Pacific      West
## West Virginia South Atlantic      South
## Wisconsin     East North Central North Central
## Wyoming       Mountain      West
```

- Añade una variable que categorice el nivel de formación (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()

```
level_of_illiteracy = cut(statex77$Illiteracy,c(0,1,2,Inf),right = FALSE,labels = c("low","some","high"))
statex77 = cbind(statex77,level_of_illiteracy)
statex77
```

```
##          Population Income Illiteracy Murder   Area
## Alabama          3615   3624         2.1   15.1  50708
## Alaska            365   6315         1.5   11.3 566432
## Arizona          2212   4530         1.8    7.8 113417
## Arkansas         2110   3378         1.9   10.1  51945
## California      21198   5114         1.1   10.3 156361
## Colorado         2541   4884         0.7    6.8 103766
## Connecticut      3100   5348         1.1    3.1   4862
## Delaware          579   4809         0.9    6.2   1982
## Florida          8277   4815         1.3   10.7  54090
```

## Georgia	4931	4091	2.0	13.9	58073
## Hawaii	868	4963	1.9	6.2	6425
## Idaho	813	4119	0.6	5.3	82677
## Illinois	11197	5107	0.9	10.3	55748
## Indiana	5313	4458	0.7	7.1	36097
## Iowa	2861	4628	0.5	2.3	55941
## Kansas	2280	4669	0.6	4.5	81787
## Kentucky	3387	3712	1.6	10.6	39650
## Louisiana	3806	3545	2.8	13.2	44930
## Maine	1058	3694	0.7	2.7	30920
## Maryland	4122	5299	0.9	8.5	9891
## Massachusetts	5814	4755	1.1	3.3	7826
## Michigan	9111	4751	0.9	11.1	56817
## Minnesota	3921	4675	0.6	2.3	79289
## Mississippi	2341	3098	2.4	12.5	47296
## Missouri	4767	4254	0.8	9.3	68995
## Montana	746	4347	0.6	5.0	145587
## Nebraska	1544	4508	0.6	2.9	76483
## Nevada	590	5149	0.5	11.5	109889
## New Hampshire	812	4281	0.7	3.3	9027
## New Jersey	7333	5237	1.1	5.2	7521
## New Mexico	1144	3601	2.2	9.7	121412
## New York	18076	4903	1.4	10.9	47831
## North Carolina	5441	3875	1.8	11.1	48798
## North Dakota	637	5087	0.8	1.4	69273
## Ohio	10735	4561	0.8	7.4	40975
## Oklahoma	2715	3983	1.1	6.4	68782
## Oregon	2284	4660	0.6	4.2	96184
## Pennsylvania	11860	4449	1.0	6.1	44966
## Rhode Island	931	4558	1.3	2.4	1049
## South Carolina	2816	3635	2.3	11.6	30225
## South Dakota	681	4167	0.5	1.7	75955
## Tennessee	4173	3821	1.7	11.0	41328
## Texas	12237	4188	2.2	12.2	262134
## Utah	1203	4022	0.6	4.5	82096
## Vermont	472	3907	0.6	5.5	9267
## Virginia	4981	4701	1.4	9.5	39780
## Washington	3559	4864	0.6	4.3	66570
## West Virginia	1799	3617	1.4	6.7	24070
## Wisconsin	4589	4468	0.7	3.0	54464
## Wyoming	376	4566	0.6	6.9	97203

##	division	region	level_of_illiteracy
## Alabama	East South Central	South	high
## Alaska	Pacific	West	some
## Arizona	Mountain	West	some
## Arkansas	West South Central	South	some
## California	Pacific	West	some
## Colorado	Mountain	West	low
## Connecticut	New England	Northeast	some
## Delaware	South Atlantic	South	low
## Florida	South Atlantic	South	some
## Georgia	South Atlantic	South	high
## Hawaii	Pacific	West	some
## Idaho	Mountain	West	low

## Illinois	East North Central	North Central	low
## Indiana	East North Central	North Central	low
## Iowa	West North Central	North Central	low
## Kansas	West North Central	North Central	low
## Kentucky	East South Central	South	some
## Louisiana	West South Central	South	high
## Maine	New England	Northeast	low
## Maryland	South Atlantic	South	low
## Massachusetts	New England	Northeast	some
## Michigan	East North Central	North Central	low
## Minnesota	West North Central	North Central	low
## Mississippi	East South Central	South	high
## Missouri	West North Central	North Central	low
## Montana	Mountain	West	low
## Nebraska	West North Central	North Central	low
## Nevada	Mountain	West	low
## New Hampshire	New England	Northeast	low
## New Jersey	Middle Atlantic	Northeast	some
## New Mexico	Mountain	West	high
## New York	Middle Atlantic	Northeast	some
## North Carolina	South Atlantic	South	some
## North Dakota	West North Central	North Central	low
## Ohio	East North Central	North Central	low
## Oklahoma	West South Central	South	some
## Oregon	Pacific	West	low
## Pennsylvania	Middle Atlantic	Northeast	some
## Rhode Island	New England	Northeast	some
## South Carolina	South Atlantic	South	high
## South Dakota	West North Central	North Central	low
## Tennessee	East South Central	South	some
## Texas	West South Central	South	high
## Utah	Mountain	West	low
## Vermont	New England	Northeast	low
## Virginia	South Atlantic	South	some
## Washington	Pacific	West	low
## West Virginia	South Atlantic	South	some
## Wisconsin	East North Central	North Central	low
## Wyoming	Mountain	West	low

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

```
west_states = statex77[grep("West",statex77$division),c("Income","level_of_illiteracy")]
west_states = west_states[order(west_states$Income,decreasing = T),]
west_states = west_states[west_states$level_of_illiteracy == "high",];
row.names(west_states[1,])
```

```
## [1] "Texas"
```

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

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

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

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

```
t(apply(df,1, sort))
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12] [,13]
## [1,]    1    1    1    1    1    1    1    2    2    2    2    2    2
## [2,]    1    1    1    1    1    1    1    1    1    1    1    1    1
## [3,]    1    1    1    1    1    1    1    1    1    1    1    2    2
## [4,]    1    1    1    1    1    1    1    1    1    1    1    1    2
```

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

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

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

```
## [14,] 5 5 5 5 5
## [15,] 5 5 5 5 5
## [16,] 5 5 5 5 5
## [17,] 5 5 5 5 5
## [18,] 5 5 5 5 5
## [19,] 5 5 5 5 5
## [20,] 5 5 5 5 5
## [21,] 5 5 5 5 5
## [22,] 5 5 5 5 5
## [23,] 5 5 5 5 5
## [24,] 5 5 5 5 5
## [25,] 5 5 5 5 5
## [26,] 5 5 5 5 5
## [27,] 5 5 5 5 5
## [28,] 5 5 5 5 5
## [29,] 5 5 5 5 5
## [30,] 5 5 5 5 5
## [31,] 5 5 5 5 5
## [32,] 5 5 5 5 5
## [33,] 5 5 5 5 5
## [34,] 5 5 5 5 5
## [35,] 5 5 5 5 5
## [36,] 5 5 5 5 5
## [37,] 5 5 5 5 5
## [38,] 5 5 5 5 5
## [39,] 5 5 5 5 5
## [40,] 5 5 5 5 5
## [41,] 4 5 5 5 5
## [42,] 5 5 5 5 5
## [43,] 5 5 5 5 5
## [44,] 5 5 5 5 5
## [45,] 5 5 5 5 5
## [46,] 5 5 5 5 5
## [47,] 5 5 5 5 5
## [48,] 5 5 5 5 5
## [49,] 5 5 5 5 5
## [50,] 4 4 5 5 5
```

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

```
t(apply(df,1, sort,decreasing=TRUE))
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12] [,13]
## [1,] 5 5 5 5 5 5 5 5 4 4 4 4 4
## [2,] 5 5 5 5 5 5 5 4 4 4 3 3 3
## [3,] 5 5 5 5 5 5 5 5 5 4 4 4 4
## [4,] 5 5 5 5 5 4 4 4 4 4 4 4 4
## [5,] 5 5 5 5 5 5 4 4 4 4 4 4 4
## [6,] 5 5 5 5 5 5 5 5 5 5 5 4 4
## [7,] 5 5 5 5 5 5 5 4 4 4 4 4 4
## [8,] 5 5 5 5 5 5 4 4 4 4 4 4 4
## [9,] 5 5 5 5 5 4 4 4 4 4 4 4 4
## [10,] 5 5 5 5 5 5 5 5 5 4 4 4 4
## [11,] 5 5 5 5 5 5 5 5 5 5 5 4 4
## [12,] 5 5 5 5 5 5 4 4 4 4 4 4 4
```

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

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

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

```
## [22,] 1 1 1 1 1
## [23,] 1 1 1 1 1
## [24,] 1 1 1 1 1
## [25,] 1 1 1 1 1
## [26,] 1 1 1 1 1
## [27,] 1 1 1 1 1
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## [29,] 1 1 1 1 1
## [30,] 1 1 1 1 1
## [31,] 1 1 1 1 1
## [32,] 1 1 1 1 1
## [33,] 1 1 1 1 1
## [34,] 1 1 1 1 1
## [35,] 1 1 1 1 1
## [36,] 1 1 1 1 1
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## [39,] 2 1 1 1 1
## [40,] 1 1 1 1 1
## [41,] 1 1 1 1 1
## [42,] 1 1 1 1 1
## [43,] 1 1 1 1 1
## [44,] 2 1 1 1 1
## [45,] 1 1 1 1 1
## [46,] 1 1 1 1 1
## [47,] 1 1 1 1 1
## [48,] 1 1 1 1 1
## [49,] 1 1 1 1 1
## [50,] 1 1 1 1 1
```

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

```
t(apply(df,1, sort,decreasing=TRUE))
```

```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12] [,13]
## [1,] 5 5 5 5 5 5 5 5 4 4 4 4 4
## [2,] 5 5 5 5 5 5 5 4 4 4 3 3 3
## [3,] 5 5 5 5 5 5 5 5 5 4 4 4 4
## [4,] 5 5 5 5 5 4 4 4 4 4 4 4 4
## [5,] 5 5 5 5 5 5 4 4 4 4 4 4 4
## [6,] 5 5 5 5 5 5 5 5 5 5 5 4 4
## [7,] 5 5 5 5 5 5 5 4 4 4 4 4 4
## [8,] 5 5 5 5 5 5 4 4 4 4 4 4 4
## [9,] 5 5 5 5 5 4 4 4 4 4 4 4 4
## [10,] 5 5 5 5 5 5 5 5 5 4 4 4 4
## [11,] 5 5 5 5 5 5 5 5 5 5 5 4 4
## [12,] 5 5 5 5 5 5 4 4 4 4 4 4 4
## [13,] 5 5 5 5 5 5 5 5 4 4 4 4 4
## [14,] 5 5 5 5 5 5 5 5 4 4 4 4 4
## [15,] 5 5 5 5 5 5 5 5 5 4 4 4 4
## [16,] 5 5 5 5 5 5 5 5 5 5 5 5 4
## [17,] 5 5 5 5 5 5 5 5 4 4 4 4 4
## [18,] 5 5 5 5 5 5 5 5 4 4 4 4 4
## [19,] 5 5 5 5 5 5 5 5 5 4 4 4 4
## [20,] 5 5 5 5 5 5 5 5 5 5 4 4 4
```

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

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

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

```
## [30,] 1 1 1 1 1
## [31,] 1 1 1 1 1
## [32,] 1 1 1 1 1
## [33,] 1 1 1 1 1
## [34,] 1 1 1 1 1
## [35,] 1 1 1 1 1
## [36,] 1 1 1 1 1
## [37,] 1 1 1 1 1
## [38,] 1 1 1 1 1
## [39,] 2 1 1 1 1
## [40,] 1 1 1 1 1
## [41,] 1 1 1 1 1
## [42,] 1 1 1 1 1
## [43,] 1 1 1 1 1
## [44,] 2 1 1 1 1
## [45,] 1 1 1 1 1
## [46,] 1 1 1 1 1
## [47,] 1 1 1 1 1
## [48,] 1 1 1 1 1
## [49,] 1 1 1 1 1
## [50,] 1 1 1 1 1
```

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 = as.data.frame(read.table("student.txt",header = TRUE),header=TRUE)
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 las 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 comando se puede utilizar para crear tablas cruzadas (cross-tabulation)

```
table(students)
```

```
## , , gender = female, population = kuopio
```

```
##
```

```
##      shoesize
```

```
## height 36 37 38 39 41 42 43 44
```

```
##      158 0 0 0 0 0 0 0 0
```

```
##      160 0 0 1 0 0 0 0 0
```

```
##      161 0 0 1 0 0 0 0 0
```

```
##      162 0 0 0 0 0 0 0 0
```

```
##      164 0 0 0 0 0 0 0 0
```

```
##      165 0 0 0 1 0 0 0 0
```

```
##      166 0 0 0 0 0 0 0 0
```

```
##      167 0 0 0 0 0 0 0 0
```

```
##      170 0 0 0 0 0 0 0 0
```

```
##      172 0 0 0 0 0 0 0 0
```

```
##      173 0 0 0 0 0 0 0 0
```

```
##      174 0 0 0 0 0 1 0 0
```

```
##      175 0 0 0 0 0 0 0 0
```

```
##      177 0 0 0 0 0 0 0 0
```

```
##      180 0 0 0 0 0 0 0 0
```

```
##      181 0 0 0 0 0 0 0 0
```

```
##
```

```
## , , gender = male, population = kuopio
```

```
##
```

```
##      shoesize
```

```
## height 36 37 38 39 41 42 43 44
```

```
##      158 0 0 0 0 0 0 0 0
```

```
##      160 0 0 0 0 0 0 0 0
```

```
##      161 0 0 0 0 0 0 0 0
```

```
##      162 0 0 0 0 0 0 0 0
```

```
##      164 0 0 0 0 0 0 0 0
```

```
##      165 0 0 0 0 0 0 0 0
```

```
##      166 0 0 0 0 0 0 0 0
```

```
##      167 0 0 0 0 0 0 0 0
```

```
##      170 0 0 0 0 0 0 1 0
```

```
##      172 0 0 0 0 0 0 1 0
```

```
##      173 0 0 0 0 0 0 0 0
```

```
##      174 0 0 0 0 0 0 0 0
```

```
##      175 0 0 0 0 0 0 0 0
```

```
##      177 0 0 0 0 0 0 0 0
```

```
##      180 0 0 0 0 0 0 0 0
```

```
##      181 0 0 0 0 0 0 0 1
```



```
##
## , , gender = female, population = tampere
##
##      shoesize
## height 36 37 38 39 41 42 43 44
##    158  1  0  0  0  0  0  0  0
##    160  0  0  0  0  0  0  0  0
##    161  0  0  0  0  0  0  0  0
##    162  0  1  0  0  0  0  0  0
##    164  0  0  0  1  0  0  0  0
##    165  0  0  0  0  0  0  0  0
##    166  0  0  1  0  0  0  0  0
##    167  0  0  1  0  0  0  0  0
##    170  0  0  0  0  0  0  0  0
##    172  0  0  0  0  0  0  0  0
##    173  0  0  0  0  0  0  0  0
##    174  0  0  0  0  0  0  0  0
##    175  0  0  0  0  0  0  0  0
##    177  0  0  0  0  0  0  0  0
##    180  0  0  0  0  0  0  0  0
##    181  0  0  0  0  0  0  0  0
##
## , , gender = male, population = tampere
##
##      shoesize
## height 36 37 38 39 41 42 43 44
##    158  0  0  0  0  0  0  0  0
##    160  0  0  0  0  0  0  0  0
##    161  0  0  0  0  0  0  0  0
##    162  0  0  0  0  0  0  0  0
##    164  0  0  0  0  0  0  0  0
##    165  0  0  0  0  0  0  0  0
##    166  0  0  0  0  0  0  0  0
##    167  0  0  0  0  0  0  0  0
##    170  0  0  0  0  0  0  0  0
##    172  0  0  0  0  0  0  0  0
##    173  0  0  0  0  1  0  0  0
##    174  0  0  0  0  0  0  0  0
##    175  0  0  0  0  0  1  0  0
##    177  0  0  0  0  0  0  1  0
##    180  0  0  0  0  0  0  1  0
##    181  0  0  0  0  0  0  0  1
```

Hay 4 observaciones diferentes, dependiendo del sexo y la población, en la que se compara el tamaño del zapato y la altura de los individuos.

- 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 será “Blue” si el estudiante es de kuopio y “Red” si es de otro sitio.

```
sym = ifelse(students$gender == "male","M","F")
colours = ifelse(students$population == "kuopio","Blue","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 set de datos student. Divídelo 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
```

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

```
## [1] 2 3 6 7 8 9 10 11 12
```

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

```
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
```

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

```
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, "student_new.txt")
```

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

```
my_list <- list(name="Fred", wife="Mary", no.children=3, child.ages=c(4,7,9))
attributes(my_list)
```

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

```
names(my_list)= my_list[2]
my_list[[2]]
```

```
## [1] "Mary"
my_list$wife
```

```
## NULL
my_list[[4]][2]
```

```
## [1] 7
length(my_list[[4]])
```

```
## [1] 3
my_list$wife <- 1:12
my_list$wife <- NULL
```

La primera sentencia crea una lista con cuatro elementos. La función `attributes(x)` accede a los atributos de un objeto, en este caso devuelve los nombres de los elementos de la lista. La segunda sentencia cambia los nombres de la lista a al segundo elemento de la lista. La tercera sentencia está accediendo al segundo elemento de la lista. La cuarta sentencia devuelve null ya que el nombre “wife” ha sido eliminado de los nombres de la lista cuando se ha hecho la asignación en la segunda sentencia. La quinta línea accede al segundo elemento del cuarto elemento de la lista. La sexta sentencia devuelve el número de elementos guardados en el cuarto elemento de la lista. La séptima sentencia añade un elemento más con nombre “wife” a la lista. La última sentencia elimina el elemento añadido justamente en la línea anterior.

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 condiciones totales, transponer columnas y filas

Vamos a volver a utilizar el datasets `mtcars`.

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

```
with(mtcars,mtcars[order(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

```
with(mtcars,mtcars[order(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 para valores menores que 150
with(mtcars,mean(mpg[hp < 150]))
```

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

```
# Media para valores mayores o iguales que 150
with(mtcars,mean(mpg[hp >= 150]))
```

```
## [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("mpg","cyl","disp","hp")]
```

```
##           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

```
mtcars.new = within(mtcars,{  
  mpgClass = ifelse(mtcars$mpg < mean(mtcars$mpg),"Low","High")  
})
```

```
mtcars.new
```

```
##           mpg cyl  disp  hp drat    wt  qsec vs am gear carb  
## 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  
## Datsun 710      22.8   4 108.0  93 3.85 2.320 18.61 1  1   4    1  
## Hornet 4 Drive  21.4   6 258.0 110 3.08 3.215 19.44 1  0   3    1  
## Hornet Sportabout 18.7   8 360.0 175 3.15 3.440 17.02 0  0   3    2  
## Valiant        18.1   6 225.0 105 2.76 3.460 20.22 1  0   3    1  
## Duster 360     14.3   8 360.0 245 3.21 3.570 15.84 0  0   3    4  
## Merc 240D      24.4   4 146.7  62 3.69 3.190 20.00 1  0   4    2  
## Merc 230       22.8   4 140.8  95 3.92 3.150 22.90 1  0   4    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  
## 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  
## Fiat 128       32.4   4  78.7  66 4.08 2.200 19.47 1  1   4    1  
## Honda Civic    30.4   4  75.7  52 4.93 1.615 18.52 1  1   4    2  
## Toyota Corolla 33.9   4  71.1  65 4.22 1.835 19.90 1  1   4    1  
## Toyota Corona  21.5   4 120.1  97 3.70 2.465 20.01 1  0   3    1  
## 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  
## Camaro Z28     13.3   8 350.0 245 3.73 3.840 15.41 0  0   3    4  
## Pontiac Firebird 19.2   8 400.0 175 3.08 3.845 17.05 0  0   3    2  
## 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  
## Lotus Europa   30.4   4  95.1 113 3.77 1.513 16.90 1  1   5    2  
## Ford Pantera L  15.8   8 351.0 264 4.22 3.170 14.50 0  1   5    4  
## Ferrari Dino   19.7   6 145.0 175 3.62 2.770 15.50 0  1   5    6  
## Maserati Bora   15.0   8 301.0 335 3.54 3.570 14.60 0  1   5    8  
## Volvo 142E     21.4   4 121.0 109 4.11 2.780 18.60 1  1   4    2  
##  
##           mpgClass
```

## Mazda RX4	High
## Mazda RX4 Wag	High
## Datsun 710	High
## Hornet 4 Drive	High
## Hornet Sportabout	Low
## Valiant	Low
## Duster 360	Low
## Merc 240D	High
## Merc 230	High
## Merc 280	Low
## Merc 280C	Low
## Merc 450SE	Low
## Merc 450SL	Low
## Merc 450SLC	Low
## Cadillac Fleetwood	Low
## Lincoln Continental	Low
## Chrysler Imperial	Low
## Fiat 128	High
## Honda Civic	High
## Toyota Corolla	High
## Toyota Corona	High
## Dodge Challenger	Low
## AMC Javelin	Low
## Camaro Z28	Low
## Pontiac Firebird	Low
## Fiat X1-9	High
## Porsche 914-2	High
## Lotus Europa	High
## Ford Pantera L	Low
## Ferrari Dino	Low
## Maserati Bora	Low
## Volvo 142E	High