

Ejercicios 3 R

Alberto Armijo Ruiz

22 de octubre de 2018

1. Dataframes.

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

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 filas, 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 function 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)], ordered = 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	1	4	4	3	5	5	5	3	5	2	5	5	4	1	5	4	1	4	1	3
## 2	1	3	5	1	2	2	1	3	1	1	4	3	1	3	3	4	3	4	2	2
## 3	2	5	5	3	5	2	3	1	1	2	5	2	2	5	3	2	1	2	5	5
## 4	2	2	2	5	1	3	4	2	4	5	4	4	5	2	4	4	4	5	5	1
## 5	2	2	3	3	5	5	3	5	5	3	5	3	1	1	4	3	5	2	4	4
## 6	2	5	2	4	5	1	4	5	2	4	3	2	2	1	2	1	2	1	5	1
## 7	5	1	3	4	5	2	3	1	3	5	2	4	2	2	2	4	2	4	4	2
## 8	5	2	4	1	3	3	1	4	3	4	5	3	3	5	4	5	1	4	5	1
## 9	5	4	1	3	1	5	4	2	5	1	5	5	2	5	2	3	1	5	2	3
## 10	1	1	5	2	4	4	2	2	4	5	3	5	2	5	1	2	2	1	1	2
## 11	4	5	2	1	5	3	4	5	3	1	4	5	1	4	1	2	3	4	3	1
## 12	5	4	3	1	4	1	4	2	2	4	1	5	5	1	1	1	4	3	4	1
## 13	5	2	4	2	2	1	3	1	2	5	2	4	2	3	2	2	1	5	1	4
## 14	1	1	4	2	5	5	3	3	5	4	1	3	3	4	3	2	5	3	1	5
## 15	4	5	3	5	5	2	1	1	1	5	4	4	1	3	1	1	1	3	3	4
## 16	1	2	1	4	1	3	3	3	5	2	4	2	3	3	5	5	4	4	4	1
## 17	3	3	5	3	4	3	5	5	1	3	4	4	4	4	2	4	5	3	4	2
## 18	1	1	1	1	4	2	3	1	2	5	1	4	2	2	2	1	4	4	1	1
## 19	1	4	1	3	5	1	4	3	3	4	4	3	5	2	1	2	2	3	2	2

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

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

```
## 26  2  5
## 27  4  4
## 28  4  4
## 29  4  1
## 30  5  2
## 31  4  4
## 32  3  4
## 33  5  2
## 34  3  2
## 35  4  2
## 36  3  1
## 37  1  3
## 38  2  3
## 39  4  3
## 40  2  1
## 41  4  1
## 42  4  2
## 43  1  2
## 44  3  1
## 45  2  2
## 46  5  2
## 47  1  2
## 48  1  1
## 49  3  1
## 50  4  4
```

- 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    2    2    2    2    2    3    3
## [2,]    1    1    1    1    1    1    1    1    1    1    2    2    2
## [3,]    1    1    1    1    2    2    2    2    2    2    2    2    2
## [4,]    1    1    1    1    1    2    2    2    2    2    2    2    2
## [5,]    1    1    1    1    1    1    1    1    1    1    2    2    2
## [6,]    1    1    1    1    1    1    1    1    1    1    1    2    2
## [7,]    1    1    1    1    1    1    2    2    2    2    2    2    2
## [8,]    1    1    1    1    1    1    1    1    1    1    1    1    1
## [9,]    1    1    1    1    1    1    1    2    2    2    2    2    2
## [10,]   1    1    1    1    1    1    1    1    1    1    2    2    2
## [11,]   1    1    1    1    1    1    1    1    1    1    1    2    2
## [12,]   1    1    1    1    1    1    1    1    1    1    1    2    2
## [13,]   1    1    1    1    1    1    1    1    1    1    1    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    1    1    1    1    2
## [16,]   1    1    1    1    1    1    1    2    2    2    2    2    2
## [17,]   1    1    1    1    2    2    2    2    2    3    3    3    3
## [18,]   1    1    1    1    1    1    1    1    1    1    1    1    1
## [19,]   1    1    1    1    1    1    1    1    1    1    2    2    2
## [20,]   1    1    1    1    1    1    1    1    1    1    1    1    2
## [21,]   1    1    1    1    1    1    1    1    2    2    2    2    2
## [22,]   1    1    1    1    1    1    1    1    1    1    2    2    2
## [23,]   1    1    1    1    1    1    1    1    1    2    2    2    2
## [24,]   1    1    1    1    1    1    2    2    2    2    2    2    2
```

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

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

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

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

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

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

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

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

##	[47,]	2	2	2	2	2	2	1	1	1	1
##	[48,]	2	2	2	2	2	1	1	1	1	1
##	[49,]	3	3	2	2	2	2	2	2	1	1
##	[50,]	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,]	2	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,]	1	1	1	1	1					
##	[17,]	2	1	1	1	1					
##	[18,]	1	1	1	1	1					
##	[19,]	1	1	1	1	1					
##	[20,]	1	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,]	1	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,]	1	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 número de elementos repetidos en un vector. Es la función más básica de clustering. Cuenta el número 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 dataset `mtcars`.

- Ordena este data set de forma ascendente según su valor 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