

```
Warning message in regularize.values(x, y, ties, missing(ties), na.rm = na.rm):
"collapsing to unique 'x' values"
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Warning message in regularize.values(x, y, ties, missing(ties), na.rm = na.rm):
"collapsing to unique 'x' values"
```

Box-Percentile Plot

▼ Práctica 3 - Actividad de aplicación

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Ejemplo 3: El archivo DirectMarketing.csv contiene datos de un vendedor de marketing directo el cual vende sus productos solo a través de correos electrónicos directos. El vendedor envía catálogos a los clientes con las características de los productos y estos ordenan directamente de los catálogos. El responsable de marketing ha desarrollado registros de clientes para aprender qué hace que algunos clientes gasten más que otros. El conjunto de datos incluye n=1000 clientes y las siguientes variables: Age (Edad del cliente; Adulta/Media/Joven); Gender (Género del cliente; masculino/femenino); OwnHome (Own/Rent); Married (soltero/casado); Location (lejos/cerca, en términos de distancia al local de una tienda que vende productos similares); Salary (sueldo anual de los clientes, en dólares); Children (número de hijos 0-3); History (del volumen de compra anterior, baja/media/alta/NA, NA significa que este cliente aún no ha adquirido ningún producto); Catalogs (número de catálogos enviados); y AmountSpent (gasto en dólares). El objetivo es explicar la variable AmountSpent en términos de las características de los clientes.

```
#Abrimos el archivo de datos indicando que las observaciones están separados por ","
#Aplicamos stringAsFactor
dDirM<-read.csv("https://raw.githubusercontent.com/Emax1/R/master/DirectMarketing.csv",sep
dDirM
#Se tienen 10 variables con 1000 observaciones por cada una de ellas
```

A data.frame: 1000 × 10

Age	Gender	OwnHome	Married	Location	Salary	Children	History	Catalogs	Amou
<fct>	<fct>	<fct>	<fct>	<fct>	<int>	<int>	<fct>	<int>	
Old	Female	Own	Single	Far	47500	0	High	6	
Middle	Male	Rent	Single	Close	63600	0	High	6	
Young	Female	Rent	Single	Close	13500	0	Low	18	
Middle	Male	Own	Married	Close	85600	1	High	18	
Middle	Female	Own	Single	Close	68400	0	High	12	
Young	Male	Own	Married	Close	30400	0	Low	6	
Middle	Female	Rent	Single	Close	48100	0	Medium	12	
Middle	Male	Own	Single	Close	68400	0	High	18	
Middle	Female	Own	Married	Close	51900	3	Low	6	
Old	Male	Own	Married	Far	80700	0	NA	18	
Young	Male	Rent	Married	Close	43700	1	NA	12	
Middle	Male	Own	Married	Far	111800	3	High	18	
Middle	Female	Own	Married	Close	44100	1	Medium	24	
Middle	Male	Own	Married	Close	111400	0	High	12	
Old	Female	Own	Married	Far	110000	0	High	24	
Middle	Female	Own	Married	Far	83100	1	NA	12	
Old	Female	Own	Married	Close	82800	0	High	24	
Middle	Male	Own	Married	Close	90100	0	High	18	
Middle	Male	Rent	Single	Close	38600	1	Low	18	
Old	Female	Rent	Single	Far	14000	0	Low	12	
Young	Female	Rent	Single	Far	14300	2	Low	6	
Young	Male	Rent	Married	Close	49800	0	Medium	12	
Old	Female	Rent	Married	Far	60200	0	High	18	
Young	Male	Rent	Married	Far	43500	1	Medium	12	
Young	Female	Rent	Married	Close	41700	1	NA	6	
Middle	Female	Own	Married	Close	92300	3	Low	18	
Middle	Male	Own	Single	Close	62800	3	NA	6	
Middle	Female	Rent	Married	Far	72900	0	High	18	
Middle	Female	Rent	Married	Close	66900	2	Medium	12	
Young	Male	Rent	Single	Close	21700	1	NA	18	
:	:	:	:	:	:	:	:	:	:

Young	Female	Rent	Single	Close	25300	3	NA	6
Old	Female	Own	Married	Close	43600	0	Medium	12
Middle	Male	Own	Married	Close	83600	2	Medium	6
Middle	Female	Own	Single	Far	56200	0	High	18
Young	Male	Rent	Married	Close	45500	0	Medium	6
Middle	Male	Own	Married	Close	99300	1	High	12
Middle	Male	Own	Married	Far	108000	3	High	24
Middle	Female	Own	Married	Far	84900	1	High	24
Middle	Male	Own	Married	Far	93000	3	High	6
Middle	Female	Own	Married	Close	82000	0	High	18
Old	Female	Own	Single	Close	33000	3	NA	6
Middle	Female	Own	Single	Close	36900	2	Low	24
Middle	Female	Rent	Single	Close	33200	1	NA	6
Young	Male	Rent	Married	Close	49300	0	NA	24
Old	Male	Rent	Married	Far	66200	0	High	24
Middle	Male	Own	Married	Far	96800	2	High	18
Old	Male	Own	Single	Close	63200	0	High	24
Old	Male	Own	Married	Far	112900	0	High	24
Middle	Female	Rent	Single	Close	32100	2	Low	24
Middle	Male	Own	Married	Far	102700	1	High	18

#Las variables observadas son:

names(dDirM)

```
'Age' · 'Gender' · 'OwnHome' · 'Married' · 'Location' · 'Salary' · 'Children' · 'History' · 'Catalogs' ·
'AmountSpent'
```

#Estructura de la matriz

str(dDirM)

#Según la estructura de la matriz se tienen variables categóricas de 3 y 2 niveles además

```
'data.frame': 1000 obs. of 10 variables:
 $ Age      : Factor w/ 3 levels "Middle","Old",...: 2 1 3 1 1 3 1 1 1 2 ...
 $ Gender   : Factor w/ 2 levels "Female","Male": 1 2 1 2 1 2 1 2 1 2 ...
 $ OwnHome  : Factor w/ 2 levels "Own","Rent": 1 2 2 1 1 1 2 1 1 1 ...
 $ Married  : Factor w/ 2 levels "Married","Single": 2 2 2 1 2 1 2 2 1 1 ...
 $ Location : Factor w/ 2 levels "Close","Far": 2 1 1 1 1 1 1 1 1 2 ...
 $ Salary   : int 47500 63600 13500 85600 68400 30400 48100 68400 51900 80700 ...
 $ Children : int 0 0 0 1 0 0 0 0 3 0 ...
 $ History  : Factor w/ 3 levels "High","Low","Medium": 1 1 2 1 1 2 3 1 2 NA ...
 $ Catalogs : int 6 6 18 18 12 6 12 18 6 18 ...
 $ AmountSpent: int 755 1318 296 2436 1304 495 782 1155 158 3034 ...
```

#Hacemos un resumen de los datos observados

```
summary(dDirM)
```

```
#Encontramos 5 variables categóricas de las cuales se muestra la media por catetoría
```

```
#De las demás se observan los estadísticos descriptivos de variables continuas
```

Age	Gender	OwnHome	Married	Location
Middle:508	Female:506	Own :516	Married:502	Close:710
Old :205	Male :494	Rent:484	Single :498	Far :290
Young :287				

Salary	Children	History	Catalogs	AmountSpent
Min. : 10100	Min. :0.000	High :255	Min. : 6.00	Min. : 38.0
1st Qu.: 29975	1st Qu.:0.000	Low :230	1st Qu.: 6.00	1st Qu.: 488.2
Median : 53700	Median :1.000	Medium:212	Median :12.00	Median : 962.0
Mean : 56104	Mean :0.934	NA's :303	Mean :14.68	Mean :1216.8
3rd Qu.: 77025	3rd Qu.:2.000		3rd Qu.:18.00	3rd Qu.:1688.5
Max. :168800	Max. :3.000		Max. :24.00	Max. :6217.0

```
#Se tienen 3 categorías de edad "Age" y se espera aplicar la media (mean) a cada uno de el
```

```
#Analizamos la categoría Age
```

```
#Tabla de dDirM$Age
```

```
table(dDirM$Age)
```

```
table(dDirM$Gender)
```

```
table(dDirM$OwnHome)
```

```
table(dDirM$Married)
```

```
table(dDirM$Location)
```

```
table(dDirM$History)
```

```
table(dDirM$Children)
```

```
#Respecto a la edad se puede observar que la mitad de los clientes corresponden a la categ
```

```
#Se observa además que gran parte de los clientes son de localidades cercanas;
```

```
#Prácticamente la mitad son clientes sin hijos y a medida que aumentan el número de hijos
```

```
#Respecto a las demás categorías encontramos prácticamente proporciones similares.
```

Middle	Old	Young
508	205	287

Female	Male
506	494

Own	Rent
516	484

Married	Single
502	498

Close	Far
710	290

High	Low	Medium
255	230	212

0	1	2	3
462	267	146	125

```
#La media general de la variable de interés o variable respuesta AmountSpent
```

```
m<-mean(dDirM$AmountSpent)
```

m

1216.77

```
#Obtendremos la media y desviación estándar de la variable AmountSpent por categoría
#X - es el vector al cual se le buscan los estadísticos
#INDEX - la variable por la que se agrupan estos
#FUN - la función que se utiliza para aplicar
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Age, FUN=mean)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Age, FUN=sd)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Gender, FUN=mean)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Gender, FUN=sd)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$OwnHome, FUN=mean)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$OwnHome, FUN=sd)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Married, FUN=mean)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Married, FUN=sd)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Location, FUN=mean)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Location, FUN=sd)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Children, FUN=mean)
tapply(X=dDirM$AmountSpent, INDEX=dDirM$Children, FUN=sd)
#Según estos datos las medias más altas de consumo corresponden a:
#Categoría de edades intermedias"Middle" seguidos muy de cerca por los jóvenes; Género más
#Al aumentnar el número de hijos se reduce la media de consumo
```

Middle:	1501.69094488189	Old:	1432.12682926829	Young:	558.623693379791
Middle:	959.273124800847	Old:	1045.93002686145	Young:	450.145948977348
Female:	1025.33992094862	Male:	1412.85020242915		
Female:	910.529013389988	Male:	972.76995076405		
Own:	1543.13565891473	Rent:	868.826446280992		
Own:	1030.25047619346	Rent:	737.270349437999		
Married:	1672.06972111554	Single:	757.813253012048		
Married:	1037.68085748728	Single:	592.200791904786		
Close:	1061.68591549296	Far:	1596.45862068966		
Close:	808.456432896427	Far:	1177.12023828293		
0:	1406.64069264069	1:	1219.5393258427	2:	940.746575342466
0:	986.843367294233	1:	971.294048724052	2:	822.307064147458
	705.202047528707			3:	831.488

```
#Llamamos la librería (Hmisc) contiene funciones para análisis descriptivo
#library(Hmisc)
#Buscamos en ella la función "describe" para el análisis descriptivo de la variable Amount
describe(dDirM$AmountSpent)
```

```
dDirM$AmountSpent
  n missing distinct      Info      Mean      Gmd      .05      .10
1000      0      852         1     1217     1014    187.9    242.9
 .25      .50      .75      .90      .95
488.2    962.0   1688.5  2551.4  3052.5

lowest :   38   43   47   62   65, highest: 5503 5564 5830 5878 6217
```

```
#Usaremos el comando "summary" para visualizar la información básica de algunas variables
dDirM[,c("AmountSpent", "Age", "Location", "Children")]
```

A data.frame: 1000 × 4

AmountSpent	Age	Location	Children
<int>	<fct>	<fct>	<int>
755	Old	Far	0
1318	Middle	Close	0
296	Young	Close	0
2436	Middle	Close	1
1304	Middle	Close	0
495	Young	Close	0
782	Middle	Close	0
1155	Middle	Close	0
158	Middle	Close	3
3034	Old	Far	0
927	Young	Close	1
2065	Middle	Far	3
704	Middle	Close	1
2136	Middle	Close	0
5564	Old	Far	0
2766	Middle	Far	1
3010	Old	Close	0
1956	Middle	Close	0
542	Middle	Close	1
410	Old	Far	0
194	Young	Far	2
827	Young	Close	0
2328	Old	Far	0
808	Young	Far	1
521	Young	Close	1
655	Middle	Close	3
707	Middle	Close	3
2075	Middle	Far	0
769	Middle	Close	2
774	Young	Close	1
:	:	:	:

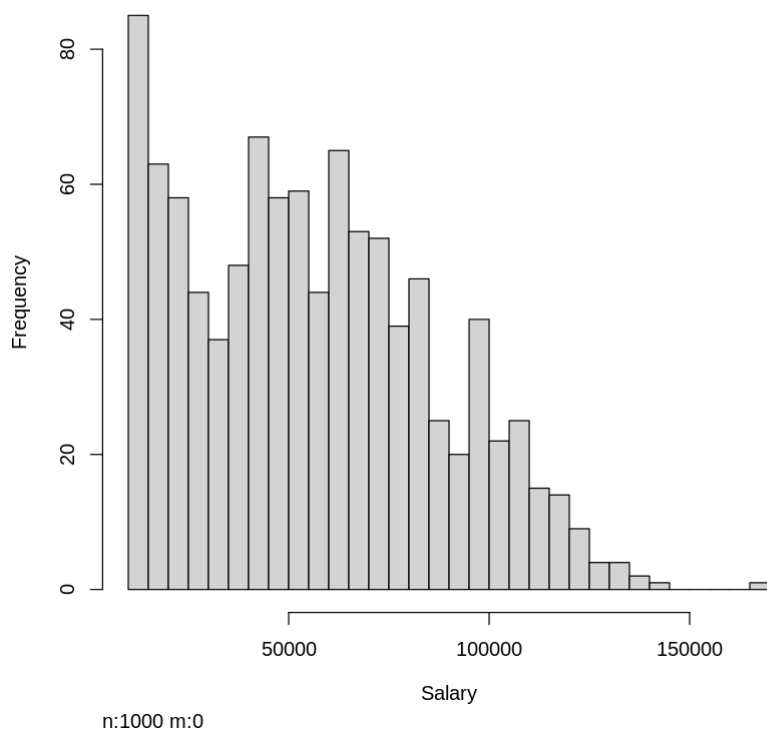
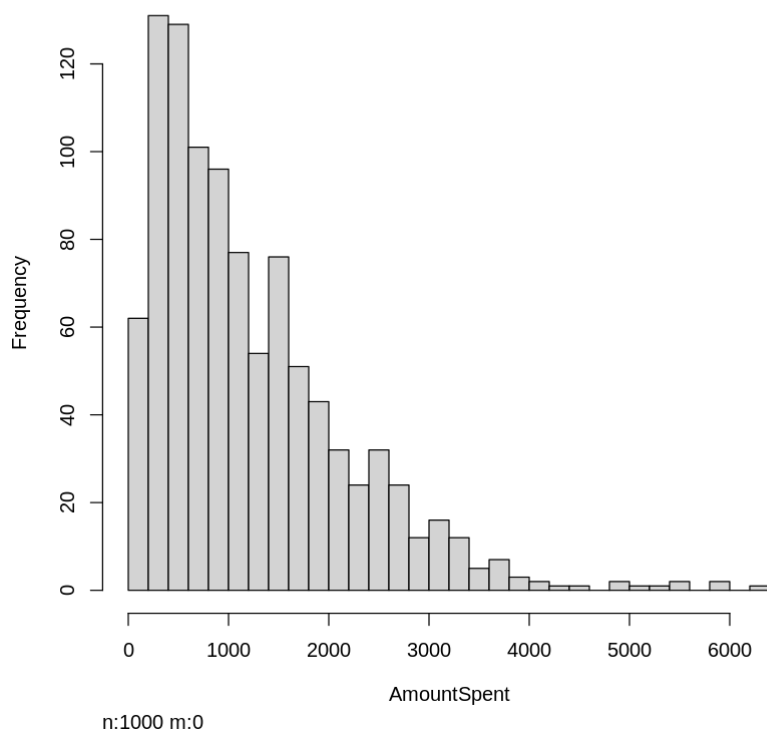
510	Young	Close	3
1047	Old	Close	0
538	Middle	Close	2
1546	Middle	Far	0
754	Young	Close	0
1504	Middle	Close	1
1849	Middle	Far	3
3072	Middle	Far	1
1046	Middle	Far	3
1800	Middle	Close	0
700	Old	Close	3
514	Middle	Close	2
320	Middle	Close	1
1692	Young	Close	0
2057	Old	Far	0
2299	Middle	Far	2
1507	Old	Close	0
6217	Old	Far	0
405	Middle	Close	2
3785	Middle	Far	1
3537	Middle	Close	2
540	Old	Far	0
941	Middle	Close	1
5503	Middle	Far	0
273	Young	Close	0
384	Young	Close	1
1073	Middle	Far	1

#Vemos un resumen de la base de datos obtenida

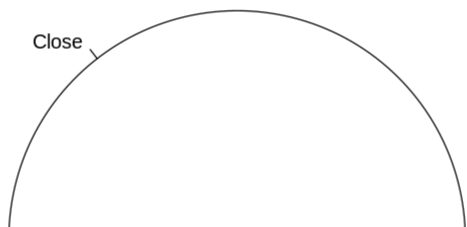
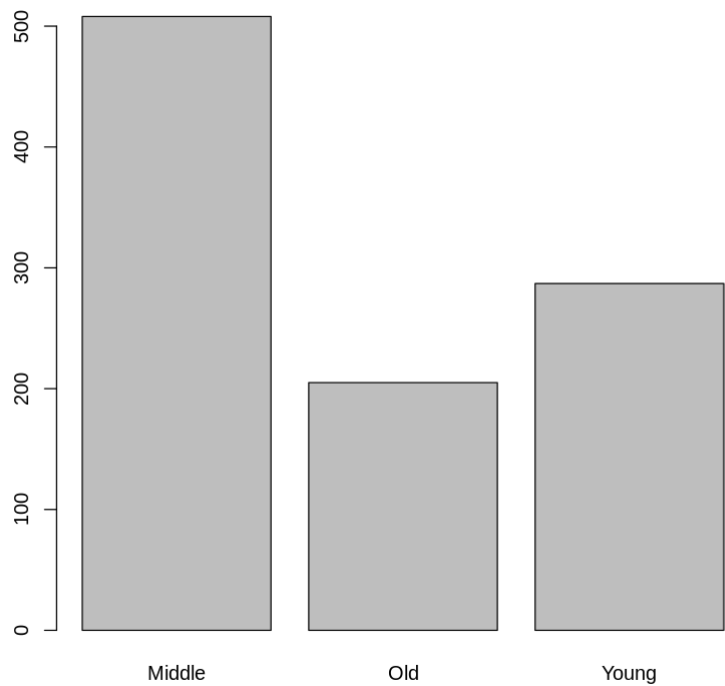
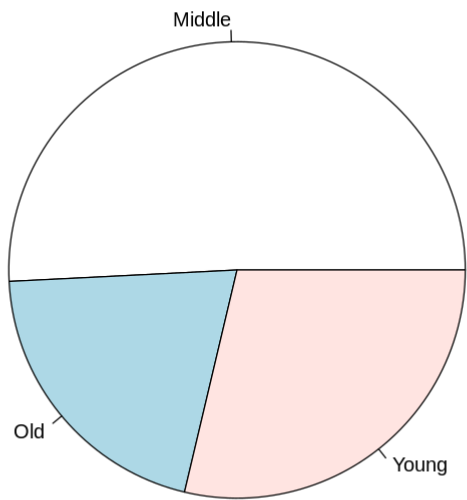
```
summary(dDirM[,c("Age", "Location", "Children", "Salary", "AmountSpent")])
```

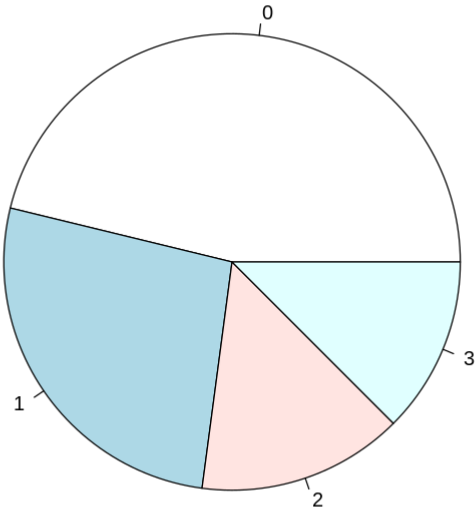
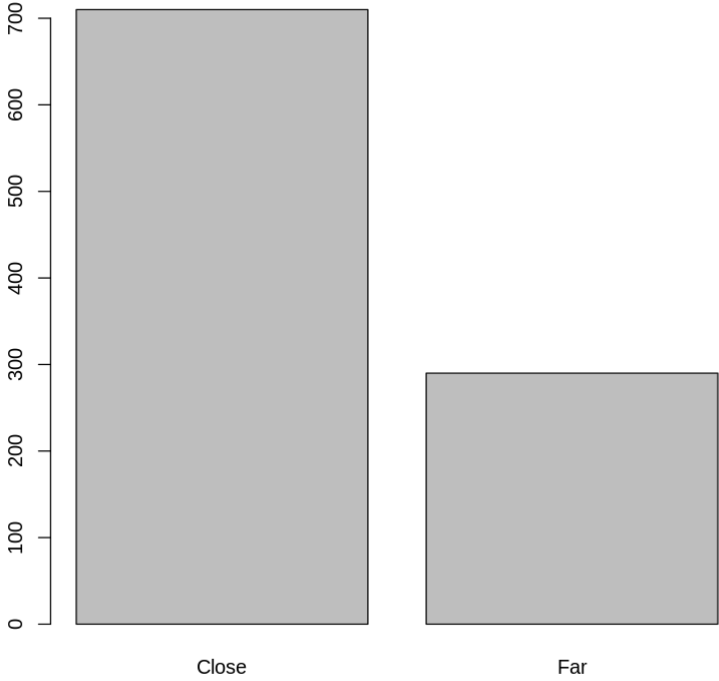
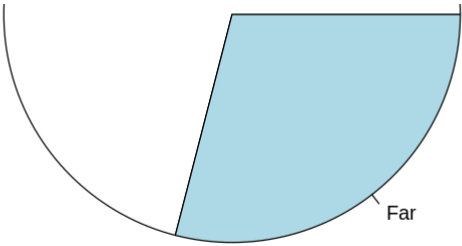
Age	Location	Children	Salary	AmountSpent
Middle:508	Close:710	Min. :0.000	Min. : 10100	Min. : 38.0
Old :205	Far :290	1st Qu.:0.000	1st Qu.: 29975	1st Qu.: 488.2
Young :287		Median :1.000	Median : 53700	Median : 962.0
		Mean :0.934	Mean : 56104	Mean :1216.8
		3rd Qu.:2.000	3rd Qu.: 77025	3rd Qu.:1688.5
		Max. :3.000	Max. :168800	Max. :6217.0

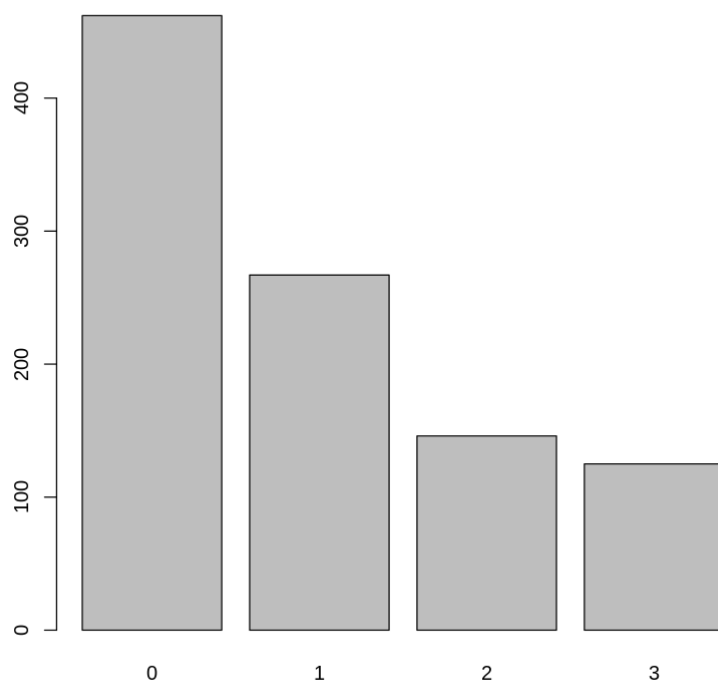
```
#Examinamos la variable "AmountSpent"
hist(dDirM["AmountSpent"]); hist(dDirM["Salary"])
```



```
#Gráfica de las tablas de datos de las variables; Gráfico de pastel y de barras
pie(table(dDirM["Age"])); barplot(table(dDirM["Age"]))
pie(table(dDirM["Location"])); barplot(table(dDirM["Location"]))
pie(table(dDirM["Children"])); barplot(table(dDirM["Children"]))
```





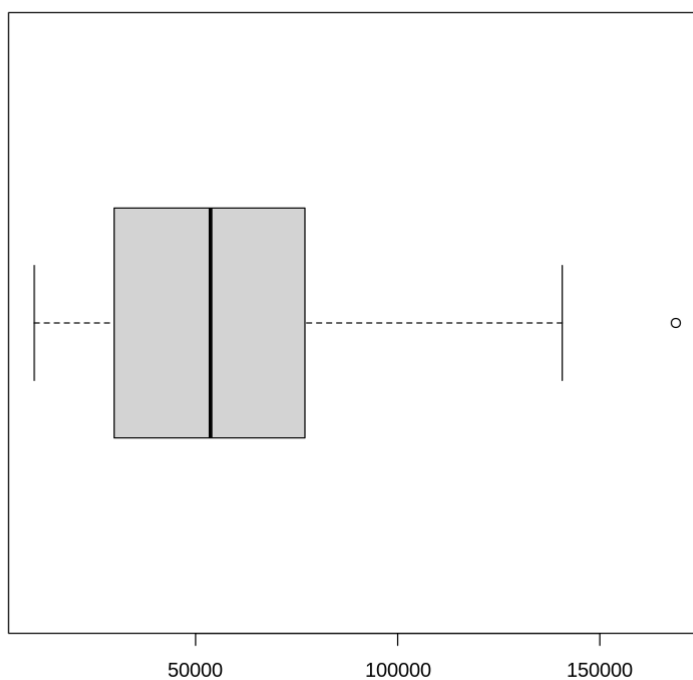
```
#Instalamos la librería "mice"  
#install.packages("mice")  
#Llamamos la librería (mice)  
#library(mice)  
#Aplicamos la función  
md.pattern(dDirM[,1:10])  
#La Fnción "md.pattern" nos permite visualizar los valores perdidos de una variable
```

A matrix: 3 × 11 of type dbl

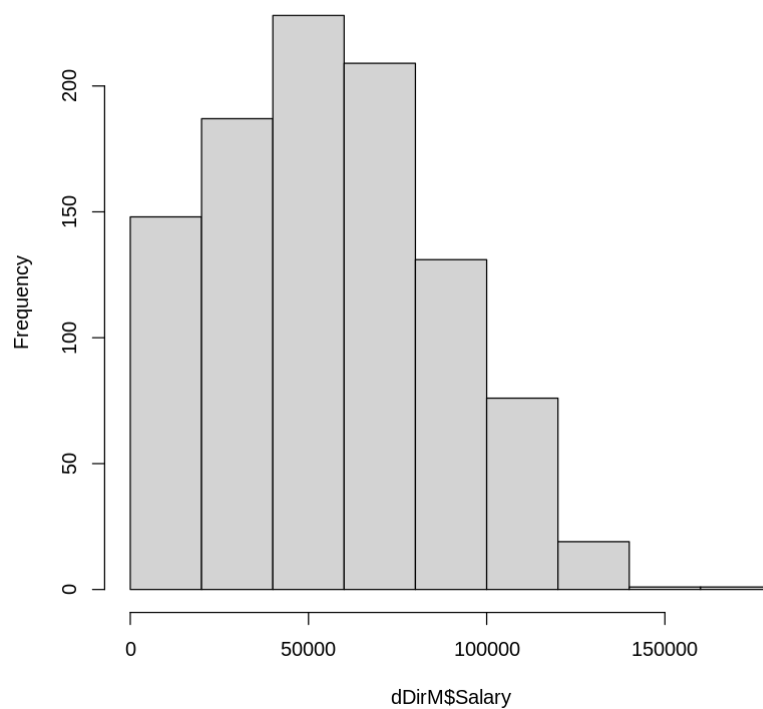
	Age	Gender	OwnHome	Married	Location	Salary	Children	Catalogs	AmountSpent
697	1	1	1	1	1	1	1	1	1
303	1	1	1	1	1	1	1	1	1
	0	0	0	0	0	0	0	0	0

```
#Hacemos un gráfico de cajas y bigotes
boxplot(dDirM$AmountSpent,horizontal=TRUE)
hist(dDirM$AmountSpent)
```

```
boxplot(dDirM$Salary, horizontal=TRUE)  
hist(dDirM$Salary)
```



Histogram of dDirM\$Salary



```
#Examinamos una nueva librería para gráficos  
#library(Hmisc)  
h3<-dDirM$AmountSpent  
hn<-h3[dDirM$AmountSpent=="No"]
```

```

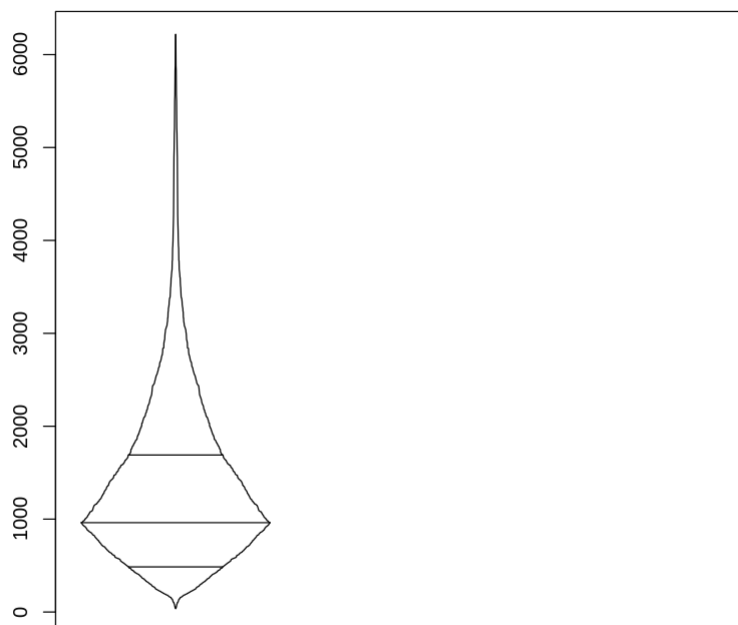
hy<-h3[dDirM$AmountSpent=="Yes"]
ds<-list(h3,hn,hy)
bpplot(ds)

Warning message in FUN(X[[i]], ...):
"no non-missing arguments to max; returning -Inf"
Warning message in FUN(X[[i]], ...):
"no non-missing arguments to max; returning -Inf"
Warning message in FUN(X[[i]], ...):
"no non-missing arguments to min; returning Inf"
Warning message in FUN(X[[i]], ...):
"no non-missing arguments to min; returning Inf"
Warning message in regularize.values(x, y, ties, missing(ties), na.rm = na.rm):
"collapsing to unique 'x' values"
Warning message in regularize.values(x, y, ties, missing(ties), na.rm = na.rm):
"collapsing to unique 'x' values"
Error in seq.default(delta, 1 - delta, delta): wrong sign in 'by' argument
Traceback:
1. bpplot(ds)
2. bpx(all.x[[i]], centers[i])
3. seq(delta, 1 - delta, delta)
4. seq.default(delta, 1 - delta, delta)
5. stop("wrong sign in 'by' argument")

```

SEARCH STACK OVERFLOW

Box-Percentile Plot



▼ Nueva sección

```
install.packages('arules')
```