



## Laboratorio 1

### Resumen del dataset

El siguiente dataset contiene 79 variables que describen las propiedades de casas en Ames, Iowa. Algunas de estas propiedades son: año de construcción, cantidad de baños, cantidad de cocinas, calidad general, condición general, tamaños de diferentes partes de la casa, entre otros. Todas ellas definen el precio de la casa. Una exploración rápida muestra lo siguiente:

```
> summary(data_training)
```

Id	MSSubClass	MSZoning	LotFrontage	LotArea	Street	Alley
Min. : 1.0	Min. : 20.0	Length:1460	Min. : 21.00	Min. : 1300	Length:1460	Length:1460
1st Qu.: 365.8	1st Qu.: 20.0	Class :character	1st Qu.: 59.00	1st Qu.: 7554	Class :character	Class :character
Median : 730.5	Median : 50.0	Mode :character	Median : 69.00	Median : 9478	Mode :character	Mode :character
Mean : 730.5	Mean : 56.9		Mean : 70.05	Mean : 10517		
3rd Qu.: 1095.2	3rd Qu.: 70.0		3rd Qu.: 80.00	3rd Qu.: 11602		
Max. : 1460.0	Max. : 190.0		Max. : 313.00	Max. : 215245		
			NA's : 259			

LotShape	LandContour	Utilities	LotConfig	LandSlope	Neighborhood
Length:1460	Length:1460	Length:1460	Length:1460	Length:1460	Length:1460
Class :character	Class :character	Class :character	Class :character	Class :character	Class :character
Mode :character	Mode :character	Mode :character	Mode :character	Mode :character	Mode :character

Condition1	Condition2	BldgType	HouseStyle	OverallQual	OverallCond	YearBuilt
Length:1460	Length:1460	Length:1460	Length:1460	Min. : 1.000	Min. : 1.000	Min. : 1872
Class :character	Class :character	Class :character	Class :character	1st Qu.: 5.000	1st Qu.: 5.000	1st Qu.: 1954
Mode :character	Mode :character	Mode :character	Mode :character	Median : 6.000	Median : 5.000	Median : 1973
				Mean : 6.099	Mean : 5.575	Mean : 1971
				3rd Qu.: 7.000	3rd Qu.: 6.000	3rd Qu.: 2000
				Max. : 10.000	Max. : 9.000	Max. : 2010

YearRemodAdd	RoofStyle	RoofMatl	Exterior1st	Exterior2nd	MasVnrType
Min. : 1950	Length:1460	Length:1460	Length:1460	Length:1460	Length:1460
1st Qu.: 1967	Class :character	Class :character	Class :character	Class :character	Class :character
Median : 1994	Mode :character	Mode :character	Mode :character	Mode :character	Mode :character
Mean : 1985					
3rd Qu.: 2004					
Max. : 2010					

MasVnrArea	ExterQual	ExterCond	Foundation	BsmtQual	BsmtCond
Min. : 0.0	Length:1460	Length:1460	Length:1460	Length:1460	Length:1460
1st Qu.: 0.0	Class :character	Class :character	Class :character	Class :character	Class :character
Median : 0.0	Mode :character	Mode :character	Mode :character	Mode :character	Mode :character
Mean : 103.7					
3rd Qu.: 166.0					
Max. : 1600.0					
NA's : 8					

BsmtExposure	BsmtFinType1	BsmtFinSF1	BsmtFinType2	BsmtFinSF2	BsmtUnfsF
Length:1460	Length:1460	Min. : 0.0	Length:1460	Min. : 0.00	Min. : 0.0
Class :character	Class :character	1st Qu.: 0.0	Class :character	1st Qu.: 0.00	1st Qu.: 223.0
Mode :character	Mode :character	Median : 383.5	Mode :character	Median : 0.00	Median : 477.5
		Mean : 443.6		Mean : 46.55	Mean : 567.2
		3rd Qu.: 712.2		3rd Qu.: 0.00	3rd Qu.: 808.0
		Max. : 5644.0		Max. : 1474.00	Max. : 2336.0

## Descripción de variables

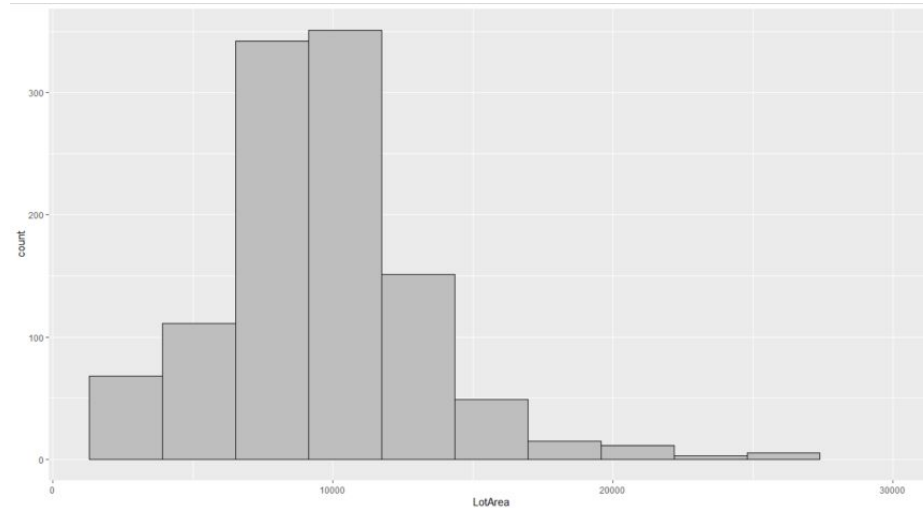
Cuantitativas		Cualitativas	
Discretas	Continuas	Ordinales	Nominales
BsmtFullBath BsmtHalfBath FullBath HalfBath Bedroom Kitchen TotRmsAbvGrd Fireplaces GarageCars EnclosedPorch 3SsnPorch ScreenPorch PoolArea	LotFrontage LotArea MasVnrArea BsmtFinSF1 BsmtFinSF2 BsmtUnfSF TotalBsmtSF 1stFlrSF 2ndFlrSF LowQualFinSF GrLivArea GarageArea WoodDeckSF OpenPorchSF MiscVal SalePrice	MSSubClass YearBuilt YearRemodAdd OverallQual OverallCond GarageYrBlt MoSold YrSold	MSZoning Street Alley LotShape LandContour Utilities LotConfig LandSlope Neighborhood Condition1 Condition2 BldgType HouseStyle RoofStyle RoofMatl Exterior1st Exterior2nd MasVnrType ExterQual ExterCond Foundation BsmtQual BsmtCond BsmtExposure BsmtFinType1 BsmtFinType2 Heating HeatingQC CentralAir Electrical KitchenQual Functional FireplaceQu GarageType GarageFinish GarageQual GarageCond PavedDrive PoolQC Fence MiscFeature SaleType SaleCondition

## Gráficos exploratorios variables numéricas

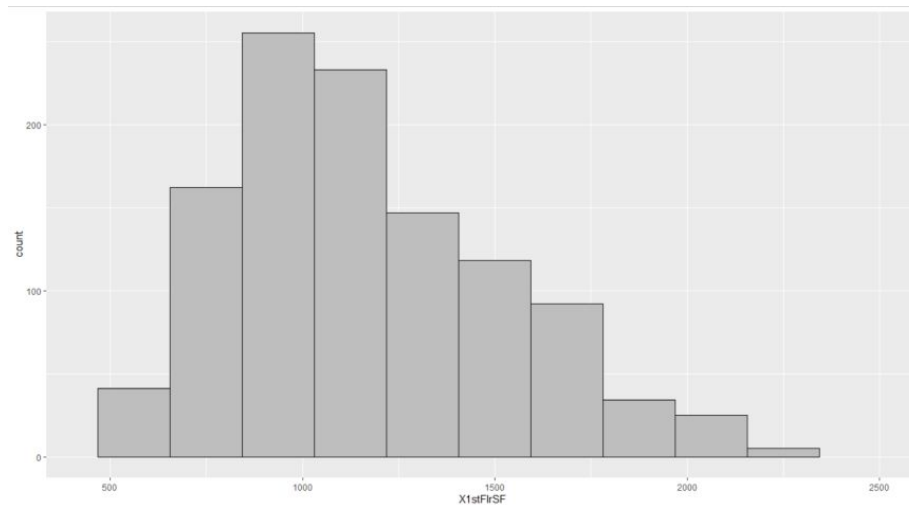
Después de generar cierta cantidad de gráficas para los datos numéricos, se pudo observar que la mayoría contiene datos atípicos. Entonces, con fines ilustrativos, no se tomaron en cuenta estos datos al limitar el rango de las gráficas. A continuación se muestran algunos ejemplos:

### 1. Histogramas

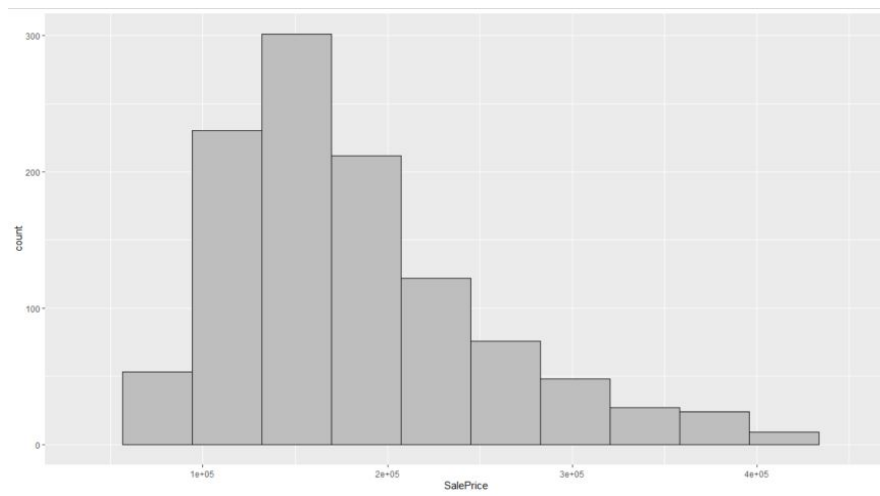
Variable: LotArea



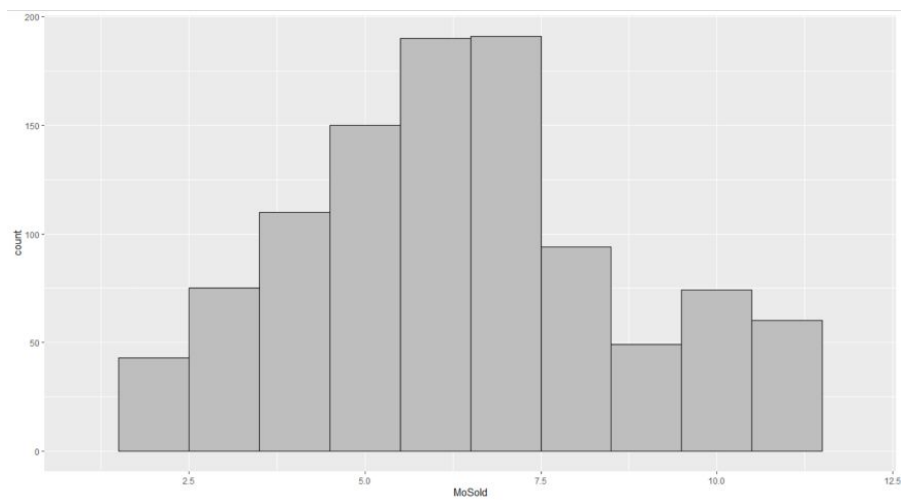
Variable: X1stFlrSF



Variable: SalePrice

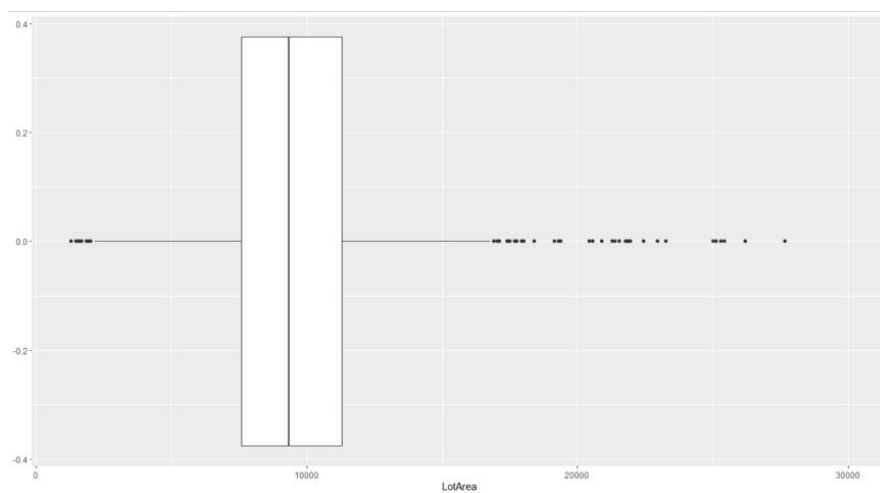


Variable: MoSold

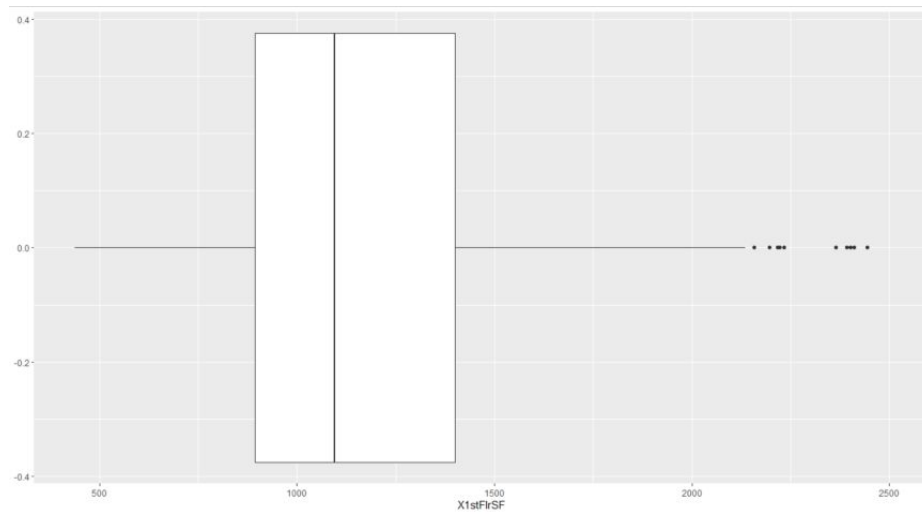


## 2. Diagrama de caja y bigotes

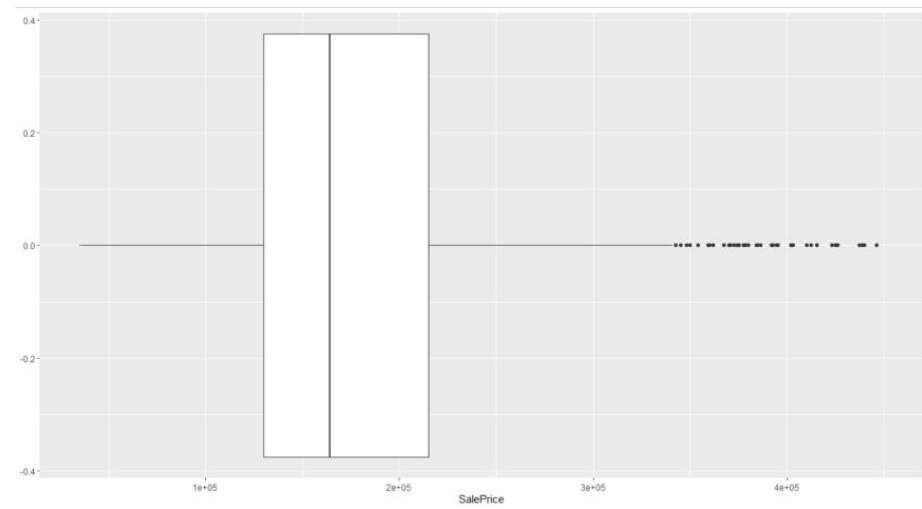
Variable: LotArea



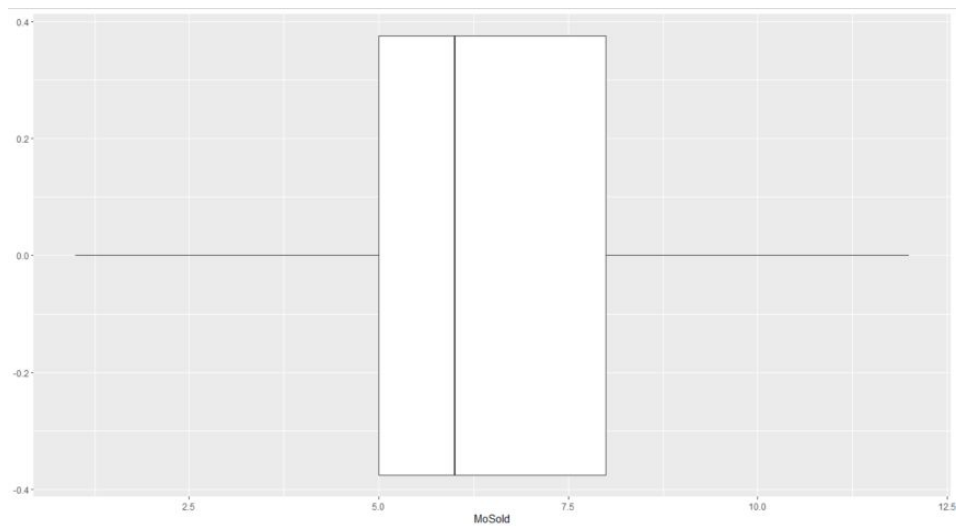
Variable: X1stFlrSF



Variable: SalePrice

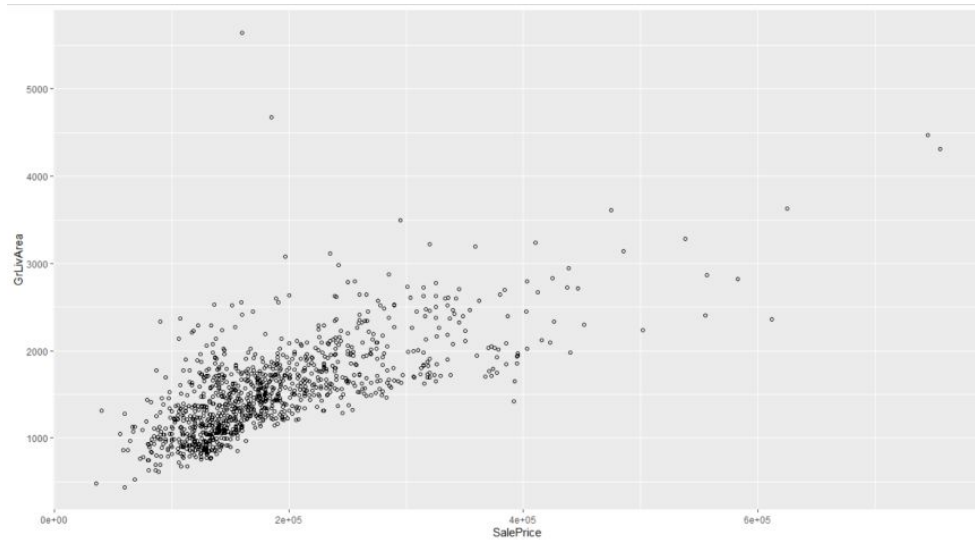


Variable: MoSold

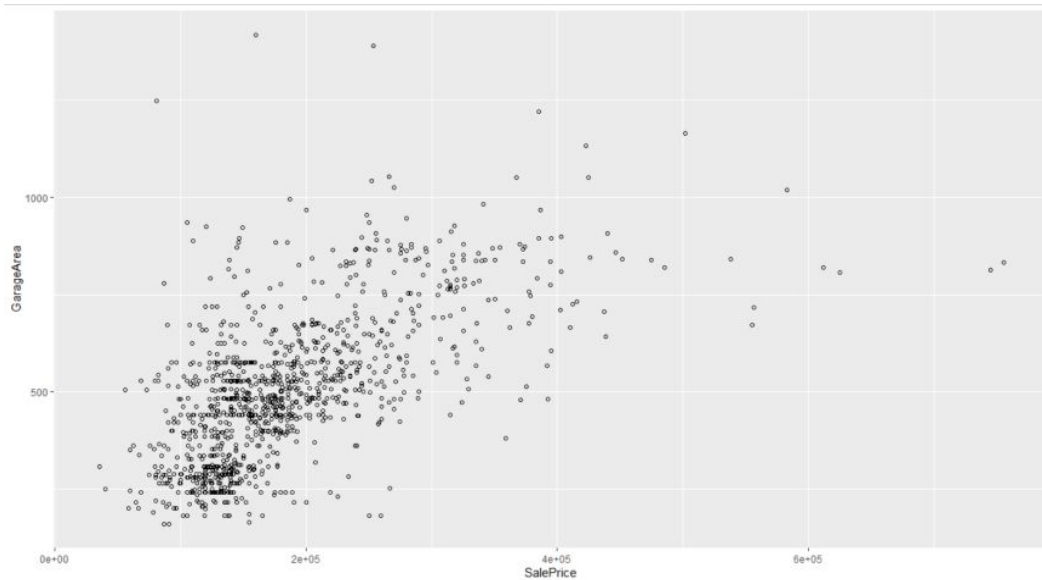


### 3. Gráfico de dispersión

Variables: GrLivArea vs SalePrice

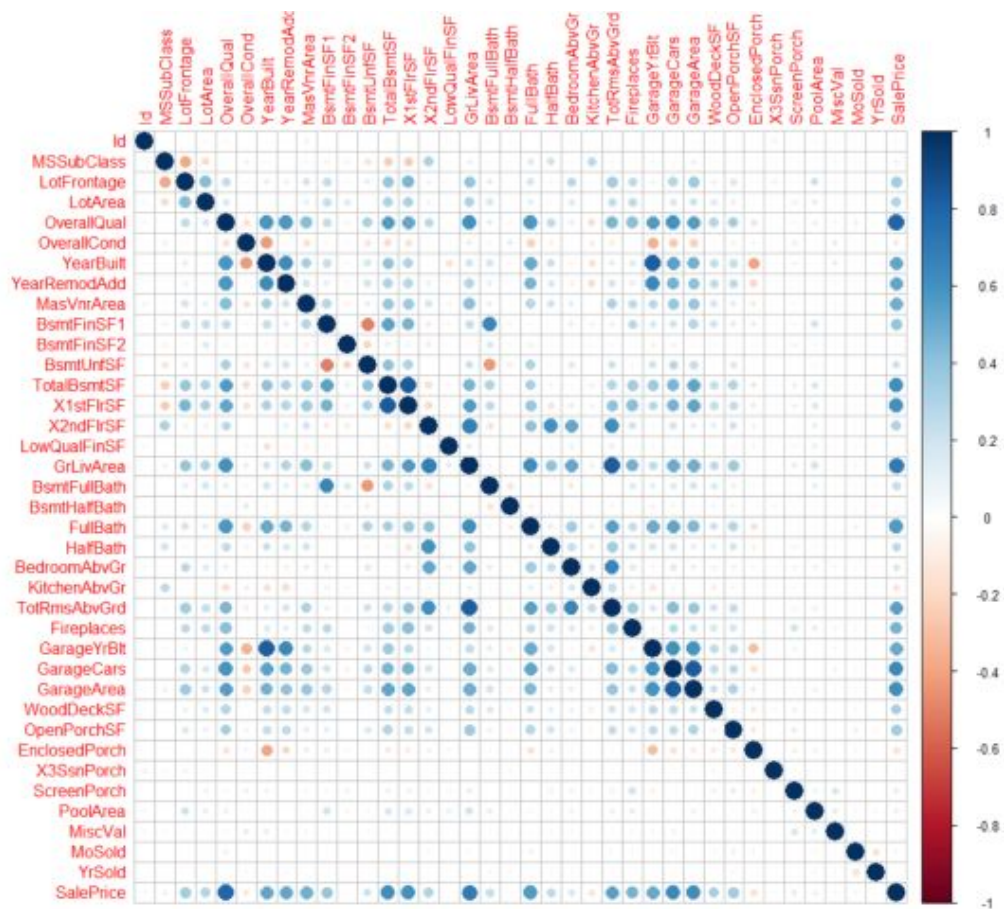


Variables: GarageArea vs SalePrice



### 4. Correlación entre variables

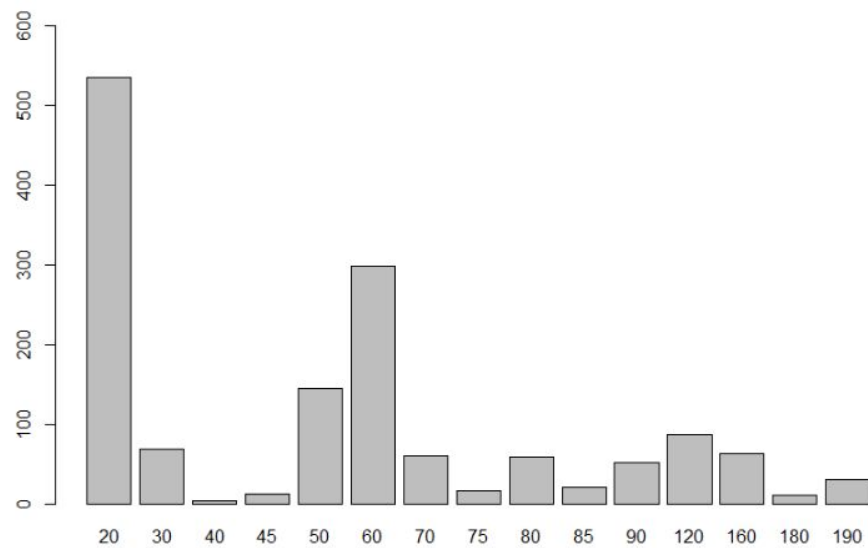
Al hacer un gráfico de correlación, se observan puntos atípicos que hacen referencia a que ciertos pares no tienen correlación. Por ejemplo, de las variables X3SsnPorch hasta YrSold, se encuentra poca o nula relación. Por otro lado, no se debería considerar el id de la casa, ya que no hay relación con alguna otra variable.



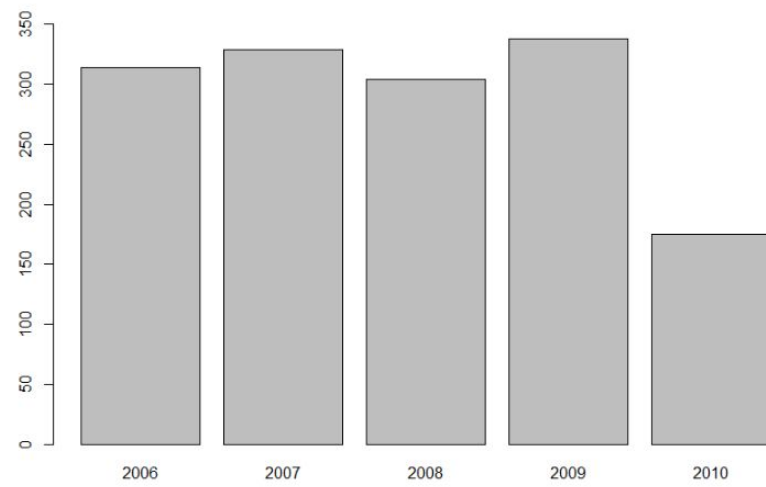
## Gráficos exploratorios variables categóricas

### 1. Gráficos de barra

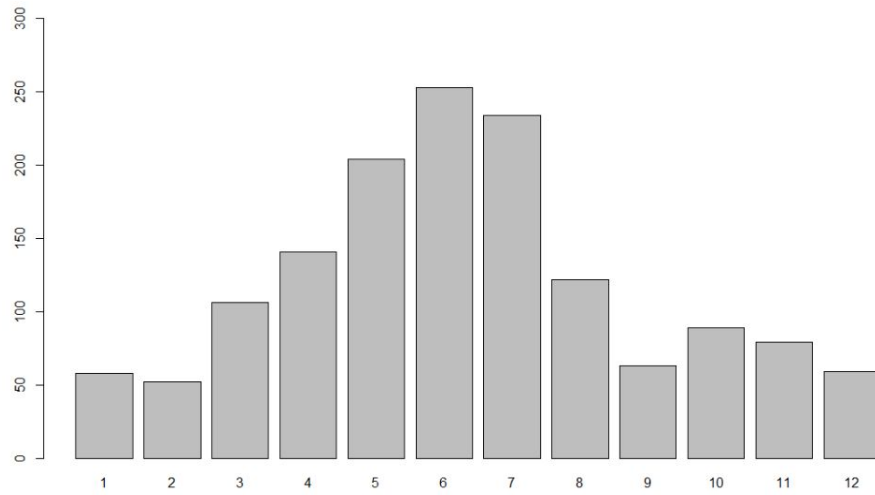
Variable: MSSubClass



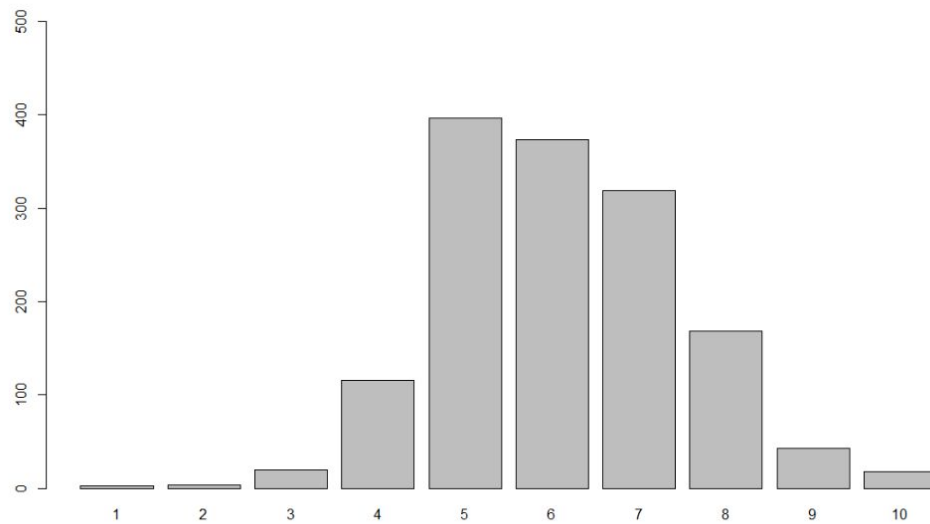
Variable: YrSold



Variable: MoSold



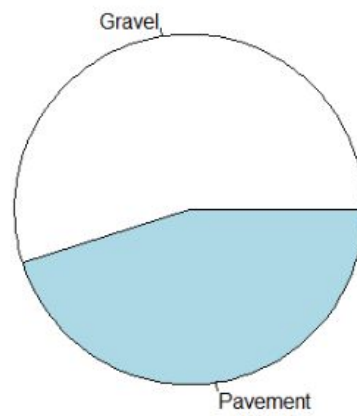
Variable: OverallQual



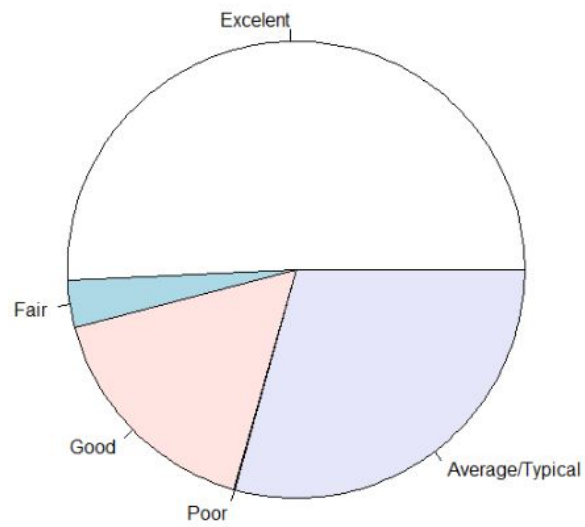


## 2. Gráficos de área (Pie)

Variable: Alley



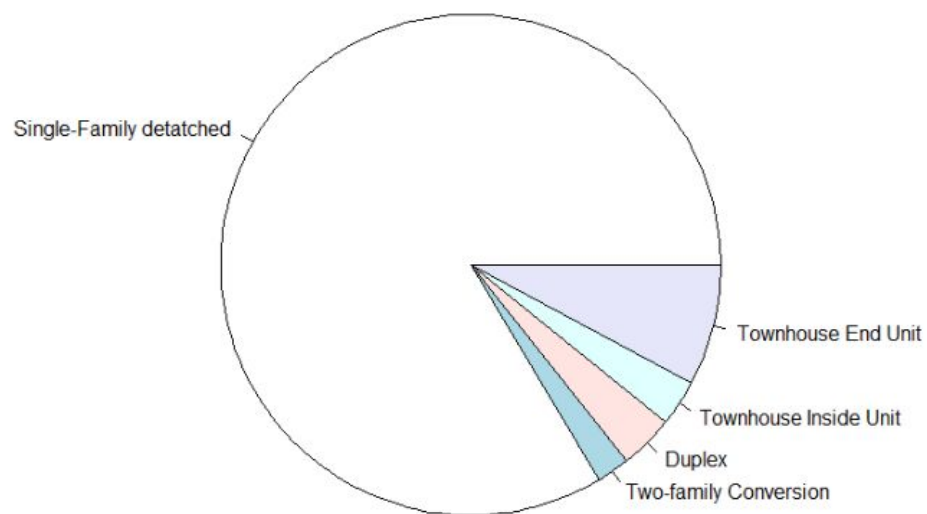
Variable: HeatingQC



Variable: BsmtQual



Variable: BldgType



### 3. Tablas de frecuencia

Variable: Utilities

AllPub	NoSeWa
1459	1

Variable: LotConfig

Corner	Cu1DSac	FR2	FR3	Inside
263	94	47	4	1052

Variable: Heating

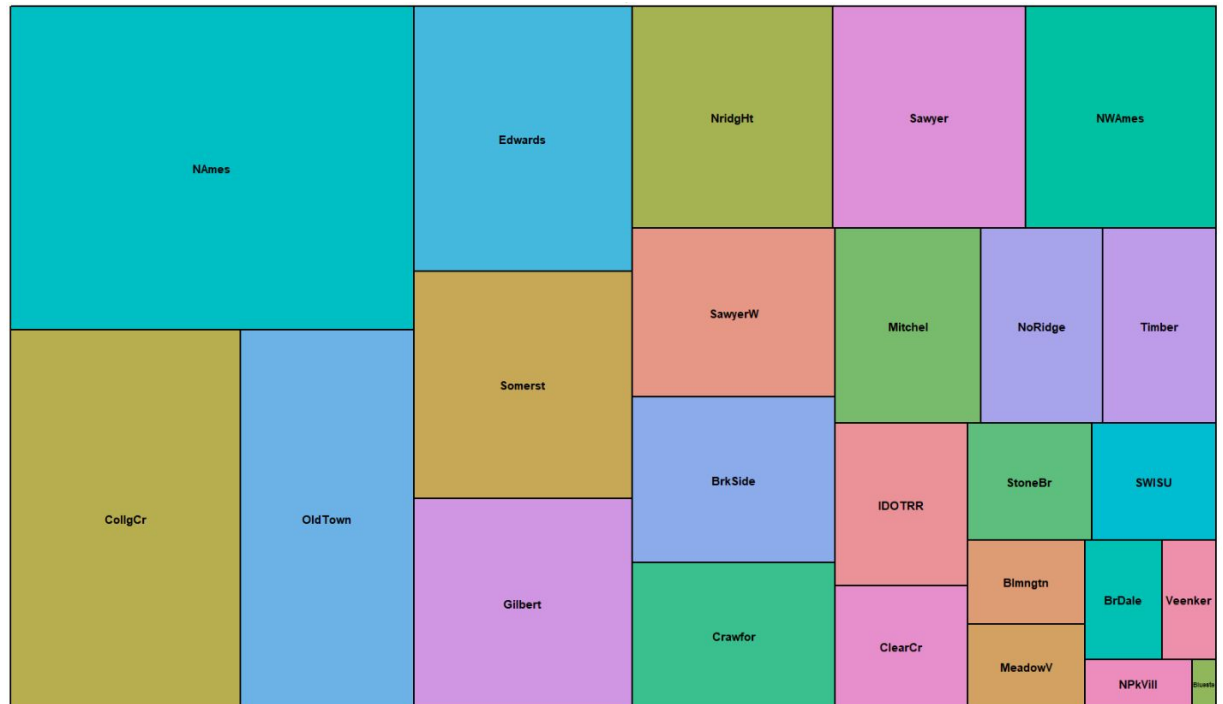
Floor	GasA	GasW	Grav	OthW	Wall
1	1428	18	7	2	4

Variable: GarageType

2Types	Attchd	Basment	BuiltIn	CarPort	Detchd
6	870	19	88	9	387

#### 4. Tablas de proporción

Variable: Neighborhood



#### Análisis de componentes principales

Primero, se analiza si es posible utilizar la técnica de análisis factorial para formar las combinaciones lineales de las variables. Se tomaron únicamente las variables más importantes. Esto se hace mediante la función “paf” y se analiza su KMO. Como el KMO es elevado (0.78), podemos decir que sí es posible utilizar esta técnica. Además, se verificó que sí vale la pena aplicar los componentes principales con el test de Bartlett. Mientras más alto sea este resultado, es mejor; en este caso se obtuvo un valor de 6359.1.

```
> pafDatos$KMO  
[1] 0.78787  
~  
> pafDatos$Bartlett  
[1] 6359.1
```

El resumen de estos resultados y la operación de paf se muestran a continuación.

```
> summary(pafDatos)
$KMO
[1] 0.78787

$MSA
      MSA
GrLivArea 0.78861
LotArea 0.85188
GarageArea 0.87015
X1stFlrSF 0.78185
YearBuilt 0.68832
YrSold 0.49347
MoSold 0.52889
OpenPorchSF 0.89640
GarageYrBlt 0.68234
LotFrontage 0.83913
TotalBsmtSF 0.77816
SalePrice 0.84857

$Bartlett
[1] 6359.1

$Communalities
      Initial Communalities Final Extraction
GrLivArea 0.599465 0.7187200
LotArea 0.230485 0.2232588
GarageArea 0.566791 0.5507416
X1stFlrSF 0.751295 0.8279618
YearBuilt 0.724289 0.7212135
YrSold 0.026933 0.0064661
MoSold 0.034857 0.0149144
OpenPorchSF 0.182229 0.1809828
GarageYrBlt 0.748883 0.9405085
LotFrontage 0.333962 0.3228366
TotalBsmtSF 0.741503 0.8199540
SalePrice 0.705067 0.7461420

$Factor.Loadings
      [,1] [,2] [,3]
GrLivArea -0.696149 -0.2916674 0.3860404
LotArea -0.341071 -0.3251006 0.0351951
GarageArea -0.730491 0.1072384 0.0749916
X1stFlrSF -0.793743 -0.3331773 -0.2948325
YearBuilt -0.615781 0.5785312 -0.0856093
YrSold 0.014425 0.0070248 -0.0787953
MoSold -0.045067 -0.0119556 0.1128733
OpenPorchSF -0.392418 0.0142645 0.1636685
GarageYrBlt -0.651410 0.7184490 0.0019896
LotFrontage -0.458674 -0.3345494 0.0230482
TotalBsmtSF -0.798741 -0.2034778 -0.3749180
SalePrice -0.840079 -0.0156998 0.2004054

$RMS
[1] 0.034568
```

## 1. Matriz de correlación

Se obtuvo la siguiente matriz de correlación. La variable más importante es SalePrice, por lo que se analizó con qué variables tenía algún tipo de relación fuerte. Las variables que afectan fuertemente al SalePrice son GrLivArea, GarageArea, X1stFlrSF y TotalBsmtSF.

```
> cor(data_training_numeric_clean_no_factors, use = "pairwise.complete.obs")
      GrLivArea LotArea GarageArea X1stFlrSF YearBuilt YrSold MoSold OpenPorchSF GarageYrBlt LotFrontage TotalBsmtSF SalePrice
GrLivArea 1.000000 0.3071635 0.487550 0.56137226 0.2049673 -0.02443609 0.0530708 0.353534 0.2437338 0.396306 0.4646447 0.705154
LotArea 0.307164 1.0000000 0.211362 0.32967869 0.0292054 -0.00690389 0.0089985 0.099170 0.0137308 0.421184 0.3025539 0.299962
GarageArea 0.487550 0.2113624 1.000000 0.52118299 0.4712859 -0.01620605 0.0375966 0.302558 0.5926352 0.356851 0.5220512 0.619330
X1stFlrSF 0.561372 0.3296787 0.521183 1.00000000 0.3088748 0.00042049 0.0277310 0.244846 0.2790531 0.451085 0.8359994 0.607969
YearBuilt 0.204967 0.0292054 0.471286 0.30887484 1.0000000 -0.00458549 0.0137844 0.235432 0.8235195 0.109726 0.4091336 0.525394
YrSold -0.024436 -0.0069039 -0.016206 0.00042049 -0.0045855 1.00000000 -0.1505766 -0.053035 0.0095961 0.013267 -0.0033775 -0.011869
MoSold 0.053071 0.0089985 0.037597 0.02773104 0.0137844 -0.1505766 1.0000000 0.089767 0.0092329 0.018815 -0.0014981 0.051568
OpenPorchSF 0.353534 0.0991700 0.302558 0.24484556 0.2354321 -0.05303516 0.0897669 1.000000 0.2571410 0.161815 0.2912859 0.343354
GarageYrBlt 0.243734 0.0137308 0.592635 0.27905312 0.8235195 0.00959605 0.0092329 0.257141 1.0000000 0.069878 0.3528768 0.504753
LotFrontage 0.396306 0.4211841 0.356851 0.45108503 0.1097256 0.01326707 0.0188145 0.161815 0.0698781 1.000000 0.3876195 0.344270
TotalBsmtSF 0.464645 0.3025539 0.522051 0.83599935 0.4091336 -0.00337749 -0.0014981 0.291286 0.3528768 0.387620 1.000000 0.615612
SalePrice 0.705154 0.2999622 0.619330 0.60796911 0.5253936 -0.01186882 0.0515681 0.343354 0.5047530 0.344270 0.6156122 1.000000
```

## 2. Componentes principales

Los componentes principales que se obtuvieron fueron los siguientes. Sin embargo, esta información es difícil de interpretar solamente con los datos normalizados. Entonces, se obtuvo la importancia de cada componente principal, en términos de desviación estándar, proporción de varianza y proporción acumulativa. En base a la

importancia, se escogerían 4 componentes principales para explicar la mayor variabilidad.

```
> compPrinc
```

```
Standard deviations (1, ..., p=12):
```

```
[1] 2.14136 1.27545 1.08630 0.93388 0.91407 0.87105 0.79046 0.76309 0.65408 0.47840 0.37539 0.36963
```

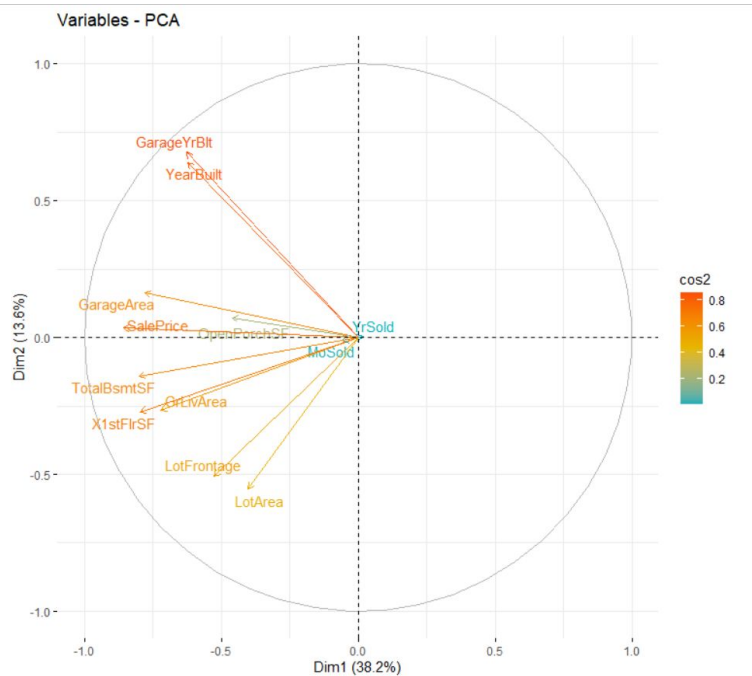
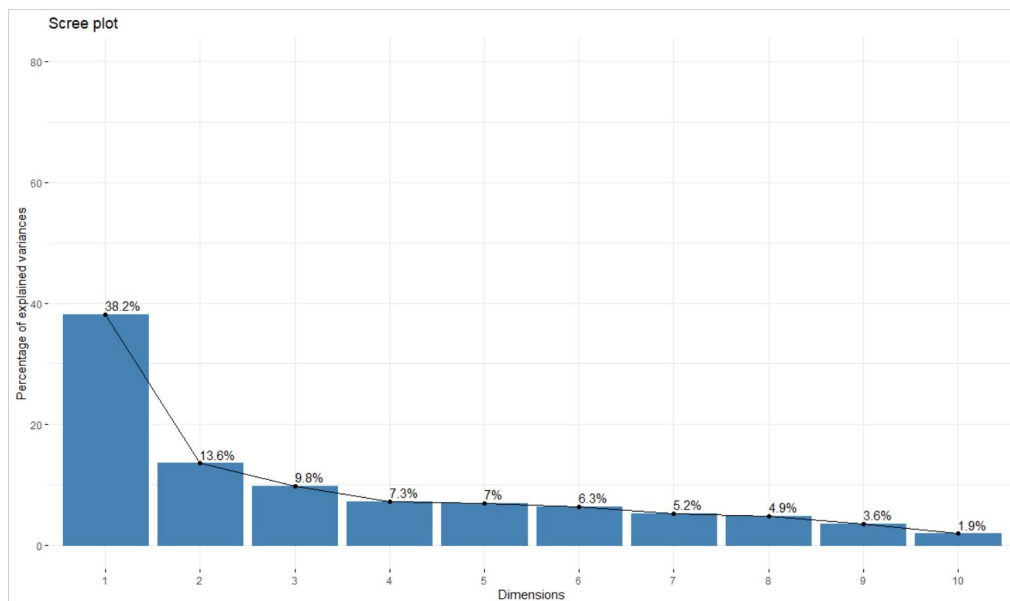
```
Rotation (n x k) = (12 x 12):
```

	PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8	PC9	PC10	PC11	PC12
GrlivArea	-0.3368438	-0.2078160	0.0796308	-0.224097	0.1963431	-0.061198	0.6163135	-0.0374082	0.2148634	-0.48340398	0.287131	-0.030995
LotArea	-0.1881431	-0.4351502	-0.0495588	0.197872	-0.2048309	0.577611	-0.0883759	-0.5724141	-0.1348264	-0.06219286	0.004549	-0.022633
GarageArea	-0.3645633	0.1268373	-0.0091185	0.040150	-0.0483596	0.069136	0.1884028	0.2838851	-0.8032795	0.10546175	0.132365	-0.228485
X1stFlrSF	-0.3721418	-0.2129309	-0.0628764	0.087953	-0.0098939	-0.460929	-0.2331255	-0.0607296	-0.0275092	-0.28891000	-0.601424	-0.296574
YearBuilt	-0.2914365	0.5005208	-0.0572864	0.162516	-0.1529395	0.155696	-0.1473556	-0.0570147	0.3957420	-0.00028082	0.249870	-0.585840
YrSold	0.0089293	0.0029695	-0.6611096	-0.629589	-0.4037966	-0.027676	-0.0137926	-0.0423107	-0.0177666	0.00182535	0.013950	-0.012316
MoSold	-0.0267071	-0.0089316	0.6831613	-0.288296	-0.6571492	-0.122753	-0.0308135	-0.0264248	-0.0012904	-0.00871842	0.021706	0.016521
OpenPorchSF	-0.2154958	0.0565545	0.2590386	-0.599776	0.5037519	0.289625	-0.4228325	-0.0186084	-0.0347424	0.02699606	-0.061429	-0.032245
GarageYrBlt	-0.2933819	0.5303384	-0.0529761	0.106318	-0.1250622	0.208039	0.0014556	0.0048883	-0.0113012	-0.35893423	-0.266385	0.599741
LotFrontage	-0.2458951	-0.3975163	-0.0595223	0.094401	-0.1662359	0.301936	-0.1467549	0.7295183	0.2910753	0.05253143	-0.017198	0.093530
TotalBsmntSF	-0.3744552	-0.1106210	-0.0801524	0.112029	0.0153291	-0.429217	-0.4029546	-0.1407805	-0.0173403	0.10743792	0.567423	0.357917
SalePrice	-0.4010260	0.0265776	0.0153141	-0.053616	0.0386528	-0.052482	0.3700704	-0.1689332	0.2152941	0.72382113	-0.279648	0.130965

```
> summary(compPrinc)
```

Importance of components:

	PC1	PC2	PC3	PC4	PC5	PC6	PC7	PC8	PC9	PC10	PC11	PC12
Standard deviation	2.141	1.275	1.0863	0.9339	0.9141	0.8710	0.7905	0.7631	0.6541	0.4784	0.3754	0.3696
Proportion of Variance	0.382	0.136	0.0983	0.0727	0.0696	0.0632	0.0521	0.0485	0.0357	0.0191	0.0117	0.0114
Cumulative Proportion	0.382	0.518	0.6160	0.6887	0.7583	0.8216	0.8736	0.9222	0.9578	0.9769	0.9886	1.0000





## Reglas de asociación

Para las reglas de asociación, se utilizó un nivel de confianza de 0.7 y un nivel de soporte de 0.2. El nivel bajo de soporte se debe a la gran cantidad de variables utilizadas. Además, solamente se tomaron en cuenta las variables que representaban una categoría definida. Se obtuvieron los siguientes resultados:

```
> reglas<-apriori(data_training_factors, parameter = list(support = 0.2,
+ confidence = 0.70,
+ target = "rules"))
Apriori
```

Parameter specification:

confidence	minval	smx	arem	aval	originalSupport	maxtime	support	minlen	maxlen	target	ext
0.7	0.1	1	none	FALSE	TRUE	5	0.2	1	10	rules	TRUE

Algorithmic control:

filter	tree	heap	memopt	load	sort	verbose
0.1	TRUE	TRUE	FALSE	TRUE	2	TRUE

Absolute minimum support count: 275

```
set item appearances ... [0 item(s)] done [0.00s].
set transactions ... [31 item(s), 1378 transaction(s)] done [0.01s].
sorting and recoding items ... [9 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 6 done [0.01s].
writing ... [174 rule(s)] done [0.00s].
creating S4 object ... done [0.00s].
```

```
> inspect(reglas)
```

lhs	rhs	support	confidence	coverage	lift	count
[1] {}	=> {RoofStyle=Gable}	0.77576	0.77576	1.00000	1.00000	1069
[2] {}	=> {SaleCondition=Normal}	0.82438	0.82438	1.00000	1.00000	1136
[3] {}	=> {LandContour=Lv1}	0.90203	0.90203	1.00000	1.00000	1243
[4] {}	=> {Electrical=SBrkr}	0.92308	0.92308	1.00000	1.00000	1272
[5] {GarageType=Detchd}	=> {LotShape=Reg}	0.22134	0.78811	0.28084	1.26281	305
[6] {GarageType=Detchd}	=> {RoofStyle=Gable}	0.23295	0.82946	0.28084	1.06922	321
[7] {GarageType=Detchd}	=> {SaleCondition=Normal}	0.24528	0.87339	0.28084	1.05944	338
[8] {GarageType=Detchd}	=> {LandContour=Lv1}	0.24673	0.87853	0.28084	0.97397	340
[9] {GarageType=Detchd}	=> {Electrical=SBrkr}	0.23295	0.82946	0.28084	0.89858	321
[10] {LotShape=IR1}	=> {GarageType=Attchd}	0.25109	0.74090	0.33890	1.17352	346
[11] {LotShape=IR1}	=> {RoofStyle=Gable}	0.26052	0.76874	0.33890	0.96143	359
[12] {LotShape=IR1}	=> {SaleCondition=Normal}	0.27939	0.82441	0.33890	1.00003	385
[13] {LotShape=IR1}	=> {LandContour=Lv1}	0.29390	0.86724	0.33890	0.96143	405
[14] {LotShape=IR1}	=> {Electrical=SBrkr}	0.32438	0.95717	0.33890	1.03694	447
[15] {LotShape=Reg}	=> {RoofStyle=Gable}	0.48766	0.78140	0.62409	1.00726	672
[16] {LotShape=Reg}	=> {SaleCondition=Normal}	0.51451	0.82442	0.62409	1.00004	709
[17] {LotShape=Reg}	=> {LandContour=Lv1}	0.58345	0.93488	0.62409	1.03642	804
[18] {LotShape=Reg}	=> {Electrical=SBrkr}	0.56241	0.90116	0.62409	0.97626	775
[19] {GarageType=Attchd}	=> {RoofStyle=Gable}	0.47750	0.75632	0.63135	0.97494	658
[20] {GarageType=Attchd}	=> {SaleCondition=Normal}	0.51669	0.81839	0.63135	0.99273	712
[21] {GarageType=Attchd}	=> {LandContour=Lv1}	0.57837	0.91609	0.63135	1.01559	797
[22] {GarageType=Attchd}	=> {Electrical=SBrkr}	0.60813	0.96322	0.63135	1.04349	838
[23] {RoofStyle=Gable}	=> {SaleCondition=Normal}	0.64224	0.82788	0.77576	1.00424	885
[24] {SaleCondition=Normal}	=> {RoofStyle=Gable}	0.64224	0.77905	0.82438	1.00424	885
[25] {RoofStyle=Gable}	=> {LandContour=Lv1}	0.70392	0.90739	0.77576	1.00594	970
[26] {LandContour=Lv1}	=> {RoofStyle=Gable}	0.70392	0.78037	0.90203	1.00594	970
[27] {RoofStyle=Gable}	=> {Electrical=SBrkr}	0.71408	0.92049	0.77576	0.99719	984
[28] {Electrical=SBrkr}	=> {RoofStyle=Gable}	0.71408	0.77358	0.92308	0.99719	984
[29] {SaleCondition=Normal}	=> {LandContour=Lv1}	0.74819	0.90757	0.82438	1.00614	1031
[30] {LandContour=Lv1}	=> {SaleCondition=Normal}	0.74819	0.82944	0.90203	1.00614	1031
[31] {SaleCondition=Normal}	=> {Electrical=SBrkr}	0.75835	0.91989	0.82438	0.99655	1045
[32] {Electrical=SBrkr}	=> {SaleCondition=Normal}	0.75835	0.82154	0.92308	0.99655	1045
[33] {LandContour=Lv1}	=> {Electrical=SBrkr}	0.83599	0.92679	0.90203	1.00402	1152
[34] {Electrical=SBrkr}	=> {LandContour=Lv1}	0.83599	0.90566	0.92308	1.00402	1152
[35] {LotShape=Reg, GarageType=Detchd}	=> {LandContour=Lv1}	0.20174	0.91148	0.22134	1.01047	278
[36] {LandContour=Lv1, GarageType=Detchd}	=> {LotShape=Reg}	0.20174	0.81765	0.24673	1.31014	278
[37] {RoofStyle=Gable, GarageType=Detchd}	=> {LandContour=Lv1}	0.20392	0.87539	0.23295	0.97046	281
[38] {LandContour=Lv1, GarageType=Detchd}	=> {RoofStyle=Gable}	0.20392	0.82647	0.24673	1.06537	281
[39] {GarageType=Detchd, SaleCondition=Normal}	=> {LandContour=Lv1}	0.21553	0.87870	0.24528	0.97413	297
[40] {LandContour=Lv1, SaleCondition=Normal}	=> {SaleCondition=Normal}	0.21553	0.87353	0.24673	1.05962	297
[41] {GarageType=Detchd, SaleCondition=Normal}	=> {Electrical=SBrkr}	0.20319	0.82840	0.24528	0.89744	280
[42] {Electrical=SBrkr, GarageType=Detchd}	=> {SaleCondition=Normal}	0.20319	0.87227	0.23295	1.05809	280
[43] {LandContour=Lv1, GarageType=Detchd}	=> {Electrical=SBrkr}	0.20464	0.82941	0.24673	0.89853	282
[44] {Electrical=SBrkr, GarageType=Detchd}	=> {LandContour=Lv1}	0.20464	0.87850	0.23295	0.97392	282
[45] {LotShape=IR1, GarageType=Attchd}	=> {SaleCondition=Normal}	0.20537	0.81792	0.25109	0.99216	283
[46] {LotShape=IR1, SaleCondition=Normal}	=> {GarageType=Attchd}	0.20537	0.73506	0.27939	1.16428	283
[47] {LotShape=IR1, GarageType=Attchd}	=> {LandContour=Lv1}	0.22061	0.87861	0.25109	0.97404	304
[48] {LotShape=IR1, LandContour=Lv1}	=> {GarageType=Attchd}	0.22061	0.75062	0.29390	1.18891	304
[49] {LotShape=IR1, GarageType=Attchd}	=> {Electrical=SBrkr}	0.24456	0.97399	0.25109	1.05515	337
[50] {LotShape=IR1, Electrical=SBrkr}	=> {GarageType=Attchd}	0.24456	0.75391	0.32438	1.19413	337
[51] {LotShape=IR1, RoofStyle=Gable}	=> {SaleCondition=Normal}	0.21843	0.83844	0.26052	1.01705	301
[52] {LotShape=IR1, SaleCondition=Normal}	=> {RoofStyle=Gable}	0.21843	0.78182	0.27939	1.00781	301
[53] {LotShape=IR1, RoofStyle=Gable}	=> {LandContour=Lv1}	0.22932	0.88022	0.26052	0.97582	316
[54] {LotShape=IR1, LandContour=Lv1}	=> {RoofStyle=Gable}	0.22932	0.78025	0.29390	1.00578	316
[55] {LotShape=IR1, RoofStyle=Gable}	=> {Electrical=SBrkr}	0.25109	0.96379	0.26052	1.04410	346
[56] {LotShape=IR1, Electrical=SBrkr}	=> {RoofStyle=Gable}	0.25109	0.77405	0.32438	0.99779	346
[57] {LotShape=IR1, SaleCondition=Normal}	=> {LandContour=Lv1}	0.24456	0.87532	0.27939	0.97039	337
[58] {LotShape=IR1, LandContour=Lv1}	=> {SaleCondition=Normal}	0.24456	0.83210	0.29390	1.00936	337
[59] {LotShape=IR1, SaleCondition=Normal}	=> {Electrical=SBrkr}	0.26705	0.95584	0.27939	1.03550	368
[60] {LotShape=IR1, Electrical=SBrkr}	=> {SaleCondition=Normal}	0.26705	0.82327	0.32438	0.99865	368



[61]	{LotShape=IRI, LandContour=Lvl}	=>	{Electrical=SBrkr}	0.28302	0.96296	0.29390	1.04321	390
[62]	{LotShape=IRI, Electrical=SBrkr}	=>	{LandContour=Lvl}	0.28302	0.87248	0.32438	0.96724	390
[63]	{LotShape=Reg, GarageType=Attchd}	=>	{RoofStyle=Gable}	0.26633	0.74593	0.35704	0.96155	367
[64]	{LotShape=Reg, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.29245	0.81911	0.35704	0.99360	403
[65]	{LotShape=Reg, GarageType=Attchd}	=>	{LandContour=Lvl}	0.34107	0.95528	0.35704	1.05904	470
[66]	{LotShape=Reg, RoofStyle=Gable}	=>	{Electrical=SBrkr}	0.34035	0.95325	0.35704	1.03269	469
[67]	{LotShape=Reg, RoofStyle=Gable}	=>	{SaleCondition=Normal}	0.39913	0.81845	0.48766	0.99281	550
[68]	{LotShape=Reg, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.39913	0.77574	0.51451	0.99997	550
[69]	{LotShape=Reg, RoofStyle=Gable}	=>	{LandContour=Lvl}	0.45501	0.93304	0.48766	1.03437	627
[70]	{LotShape=Reg, LandContour=Lvl}	=>	{RoofStyle=Gable}	0.45501	0.77985	0.58345	1.00527	627
[71]	{LotShape=Reg, RoofStyle=Gable}	=>	{Electrical=SBrkr}	0.43614	0.89435	0.48766	0.96887	601
[72]	{LotShape=Reg, Electrical=SBrkr}	=>	{RoofStyle=Gable}	0.43614	0.77548	0.56241	0.99964	601
[73]	{LotShape=Reg, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.48186	0.93653	0.51451	1.03825	664
[74]	{LotShape=Reg, LandContour=Lvl}	=>	{SaleCondition=Normal}	0.48186	0.82587	0.58345	1.00180	664
[75]	{LotShape=Reg, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.46154	0.89704	0.51451	0.97179	636
[76]	{LotShape=Reg, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.46154	0.82065	0.56241	0.99547	636
[77]	{LotShape=Reg, LandContour=Lvl}	=>	{Electrical=SBrkr}	0.52830	0.90547	0.58345	0.98093	728
[78]	{LotShape=Reg, Electrical=SBrkr}	=>	{LandContour=Lvl}	0.52830	0.93935	0.56241	1.04138	728
[79]	{RoofStyle=Gable, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.39332	0.82371	0.47750	0.99918	542
[80]	{GarageType=Attchd, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.39332	0.76124	0.51669	0.98128	542
[81]	{RoofStyle=Gable, GarageType=Attchd}	=>	{LandContour=Lvl}	0.44194	0.92553	0.47750	1.02605	609
[82]	{LandContour=Lvl, GarageType=Attchd}	=>	{RoofStyle=Gable}	0.44194	0.76412	0.57837	0.98499	609
[83]	{RoofStyle=Gable, GarageType=Attchd}	=>	{Electrical=SBrkr}	0.45936	0.96201	0.47750	1.04217	633
[84]	{Electrical=SBrkr, GarageType=Attchd}	=>	{RoofStyle=Gable}	0.45936	0.75537	0.60813	0.97371	633
[85]	{GarageType=Attchd, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.47605	0.92135	0.51669	1.02141	656
[86]	{LandContour=Lvl, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.47605	0.82309	0.57837	0.99843	656
[87]	{GarageType=Attchd, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.49492	0.95787	0.51669	1.03769	682
[88]	{Electrical=SBrkr, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.49492	0.81384	0.60813	0.98721	682
[89]	{LandContour=Lvl, GarageType=Attchd}	=>	{Electrical=SBrkr}	0.55878	0.96612	0.57837	1.04663	770
[90]	{Electrical=SBrkr, GarageType=Attchd}	=>	{LandContour=Lvl}	0.55878	0.91885	0.60813	1.01865	770
[91]	{RoofStyle=Gable, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.58418	0.90960	0.64224	1.00840	805
[92]	{LandContour=Lvl, RoofStyle=Gable}	=>	{SaleCondition=Normal}	0.58418	0.82990	0.70392	1.00669	805
[93]	{LandContour=Lvl, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.58418	0.78080	0.74819	1.00649	805
[94]	{RoofStyle=Gable, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.58999	0.91864	0.64224	0.99520	813
[95]	{RoofStyle=Gable, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.58999	0.82622	0.71408	1.00223	813
[96]	{Electrical=SBrkr, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.58999	0.77799	0.75835	1.00287	813
[97]	{LandContour=Lvl, RoofStyle=Gable}	=>	{Electrical=SBrkr}	0.64949	0.92268	0.70392	0.99957	895
[98]	{RoofStyle=Gable, Electrical=SBrkr}	=>	{LandContour=Lvl}	0.64949	0.90955	0.71408	1.00834	895
[99]	{LandContour=Lvl, Electrical=SBrkr}	=>	{RoofStyle=Gable}	0.64949	0.77691	0.83599	1.00148	895
[100]	{LandContour=Lvl, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.69231	0.92532	0.74819	1.00242	954
[101]	{Electrical=SBrkr, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.69231	0.91292	0.75835	1.01207	954
[102]	{LandContour=Lvl, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.69231	0.82812	0.83599	1.00454	954
[103]	{LotShape=IRI, LandContour=Lvl, GarageType=Attchd}	=>	{Electrical=SBrkr}	0.21480	0.97368	0.22061	1.05482	296
[104]	{LotShape=IRI, Electrical=SBrkr, GarageType=Attchd}	=>	{LandContour=Lvl}	0.21480	0.87834	0.24456	0.97373	296
[105]	{LotShape=IRI, LandContour=Lvl, Electrical=SBrkr}	=>	{GarageType=Attchd}	0.21480	0.75897	0.28302	1.02015	296
[106]	{LotShape=IRI, RoofStyle=Gable, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.21045	0.96346	0.21843	1.04374	290
[107]	{LotShape=IRI, RoofStyle=Gable, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.21045	0.83815	0.25109	1.01670	290
[108]	{LotShape=IRI, Electrical=SBrkr, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.21045	0.78804	0.26705	1.01583	290
[109]	{LotShape=IRI, LandContour=Lvl, RoofStyle=Gable}	=>	{Electrical=SBrkr}	0.22206	0.96835	0.22932	1.04905	306
[110]	{LotShape=IRI, RoofStyle=Gable, Electrical=SBrkr}	=>	{LandContour=Lvl}	0.22206	0.88439	0.25109	0.98405	306
[111]	{LotShape=IRI, LandContour=Lvl, Electrical=SBrkr}	=>	{RoofStyle=Gable}	0.22206	0.78462	0.28302	1.01141	306
[112]	{LotShape=IRI, LandContour=Lvl, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.23585	0.96439	0.24456	1.04476	325
[113]	{LotShape=IRI, Electrical=SBrkr, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.23585	0.88315	0.26705	0.97907	325
[114]	{LotShape=IRI, LandContour=Lvl, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.23585	0.83333	0.28302	1.01086	325
[115]	{LotShape=Reg, RoofStyle=Gable, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.21626	0.81199	0.26633	0.98497	298
[116]	{LotShape=Reg, GarageType=Attchd, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.21626	0.73945	0.29245	0.95320	298
[117]	{LotShape=Reg, RoofStyle=Gable, GarageType=Attchd}	=>	{LandContour=Lvl}	0.25399	0.95368	0.26633	1.05726	350
[118]	{LotShape=Reg, LandContour=Lvl, GarageType=Attchd}	=>	{RoofStyle=Gable}	0.25399	0.74468	0.34107	0.95993	350
[119]	{LotShape=Reg, RoofStyle=Gable, GarageType=Attchd}	=>	{Electrical=SBrkr}	0.25254	0.94823	0.26633	1.02725	348
[120]	{LotShape=Reg, Electrical=SBrkr, GarageType=Attchd}	=>	{RoofStyle=Gable}	0.25254	0.74200	0.34035	0.95648	348
[121]	{LotShape=Reg, GarageType=Attchd, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.27866	0.95285	0.29245	1.05634	384
[122]	{LotShape=Reg, LandContour=Lvl, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.27866	0.81702	0.34107	0.99107	384
[123]	{LotShape=Reg, GarageType=Attchd, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.27721	0.94789	0.29245	1.02688	382
[124]	{LotShape=Reg, Electrical=SBrkr, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.27721	0.81450	0.34035	0.98801	382
[125]	{LotShape=Reg, LandContour=Lvl, GarageType=Attchd}	=>	{Electrical=SBrkr}	0.32729	0.95957	0.34107	1.03954	451
[126]	{LotShape=Reg, Electrical=SBrkr, GarageType=Attchd}	=>	{LandContour=Lvl}	0.32729	0.96162	0.34035	1.06606	451
[127]	{LotShape=Reg, RoofStyle=Gable, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.37228	0.93273	0.39913	1.03403	513
[128]	{LotShape=Reg, LandContour=Lvl, RoofStyle=Gable}	=>	{SaleCondition=Normal}	0.37228	0.81818	0.45501	0.99248	513
[129]	{LotShape=Reg, LandContour=Lvl, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.37228	0.77259	0.48186	0.99591	513
[130]	{LotShape=Reg, RoofStyle=Gable, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.35559	0.89091	0.39913	0.96515	490
[131]	{LotShape=Reg, RoofStyle=Gable, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.35559	0.81531	0.43614	0.98899	490
[132]	{LotShape=Reg, Electrical=SBrkr, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.35559	0.77044	0.46154	0.99314	490
[133]	{LotShape=Reg, LandContour=Lvl, RoofStyle=Gable}	=>	{Electrical=SBrkr}	0.40784	0.89633	0.45501	0.97103	562
[134]	{LotShape=Reg, RoofStyle=Gable, Electrical=SBrkr}	=>	{LandContour=Lvl}	0.40784	0.93511	0.43614	1.03667	562
[135]	{LotShape=Reg, LandContour=Lvl, Electrical=SBrkr}	=>	{RoofStyle=Gable}	0.40784	0.77198	0.52830	0.99512	562
[136]	{LotShape=Reg, LandContour=Lvl, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.43469	0.90211	0.48186	0.97728	599
[137]	{LotShape=Reg, Electrical=SBrkr, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.43469	0.94182	0.46154	1.04411	599
[138]	{LotShape=Reg, LandContour=Lvl, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.43469	0.82280	0.52830	0.99808	599
[139]	{RoofStyle=Gable, GarageType=Attchd, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.36575	0.92989	0.39332	1.03088	504
[140]	{LandContour=Lvl, RoofStyle=Gable, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.36575	0.82759	0.44194	1.00389	504
[141]	{LandContour=Lvl, GarageType=Attchd, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.36575	0.76829	0.47605	0.99037	504
[142]	{RoofStyle=Gable, GarageType=Attchd, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.37663	0.95756	0.39332	1.03736	519
[143]	{RoofStyle=Gable, Electrical=SBrkr, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.37663	0.81991	0.45936	0.99457	519
[144]	{Electrical=SBrkr, GarageType=Attchd, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.37663	0.76100	0.49492	0.98097	519
[145]	{LandContour=Lvl, RoofStyle=Gable, GarageType=Attchd}	=>	{Electrical=SBrkr}	0.42598	0.96388	0.44194	1.04420	587
[146]	{RoofStyle=Gable, Electrical=SBrkr, GarageType=Attchd}	=>	{LandContour=Lvl}	0.42598	0.92733	0.45936	1.02805	587
[147]	{LandContour=Lvl, Electrical=SBrkr, GarageType=Attchd}	=>	{RoofStyle=Gable}	0.42598	0.76234	0.55878	0.98270	587
[148]	{LotShape=Reg, LandContour=Lvl, GarageType=Attchd, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.45791	0.96189	0.47605	1.04205	631
[149]	{Electrical=SBrkr, GarageType=Attchd, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.45791	0.92522	0.49492	1.02571	631
[150]	{LandContour=Lvl, Electrical=SBrkr, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.45791	0.81948	0.55878	0.99405	631
[151]	{LandContour=Lvl, RoofStyle=Gable, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.53919	0.92298	0.58418	0.99990	743
[152]	{RoofStyle=Gable, Electrical=SBrkr, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.53919	0.91390	0.58999	1.01316	743
[153]	{LandContour=Lvl, RoofStyle=Gable, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.53919	0.83017	0.64949	1.00702	743
[154]	{LandContour=Lvl, Electrical=SBrkr, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.53919	0.77883	0.69231	1.00395	743
[155]	{LotShape=Reg, RoofStyle=Gable, GarageType=Attchd, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.20610	0.95302	0.21626	1.05653	284
[156]	{LotShape=Reg, LandContour=Lvl, RoofStyle=Gable, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.20610	0.81143	0.25399	0.98429	284
[157]	{LotShape=Reg, LandContour=Lvl, GarageType=Attchd, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.20610	0.73958	0.27866	0.95336	284
[158]	{LotShape=Reg, RoofStyle=Gable, GarageType=Attchd, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.20392	0.94295	0.21626	1.02153	281
[159]	{LotShape=Reg, RoofStyle=Gable, Electrical=SBrkr, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.20392	0.80747	0.25254	0.97949	281
[160]	{LotShape=Reg, Electrical=SBrkr, GarageType=Attchd, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.20392	0.73560	0.27721	0.94823	281
[161]	{LotShape=Reg, LandContour=Lvl, RoofStyle=Gable, GarageType=Attchd}	=>	{Electrical=SBrkr}	0.24165	0.95143	0.25399	1.03071	333
[162]	{LotShape=Reg, RoofStyle=Gable, Electrical=SBrkr, GarageType=Attchd}	=>	{LandContour=Lvl}	0.24165	0.95690	0.25254	1.06082	333
[163]	{LotShape=Reg, LandContour=Lvl, Electrical=SBrkr, GarageType=Attchd}	=>	{RoofStyle=Gable}	0.24165	0.73836	0.32729	0.95179	333
[164]	{LotShape=Reg, LandContour=Lvl, GarageType=Attchd, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.26633	0.95573	0.27866	1.03537	367
[165]	{LotShape=Reg, Electrical=SBrkr, GarageType=Attchd, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.26633	0.96073	0.27721	1.06508	367
[166]	{LotShape=Reg, LandContour=Lvl, Electrical=SBrkr, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.26633	0.81375	0.32729	0.98710	367
[167]	{LotShape=Reg, LandContour=Lvl, RoofStyle=Gable, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.33309	0.89474	0.37228	0.96930	459
[168]	{LotShape=Reg, RoofStyle=Gable, Electrical=SBrkr, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.33309	0.93673	0.35559	1.03847	459
[169]	{LotShape=Reg, LandContour=Lvl, RoofStyle=Gable, Electrical=SBrkr}	=>	{SaleCondition=Normal}	0.33309	0.81673	0.40784	0.99071	459
[170]	{LotShape=Reg, LandContour=Lvl, Electrical=SBrkr, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.33309	0.76628	0.43469	0.98777	459
[171]	{LandContour=Lvl, RoofStyle=Gable, GarageType=Attchd, SaleCondition=Normal}	=>	{Electrical=SBrkr}	0.35123	0.96032	0.36575	1.04034	484
[172]	{RoofStyle=Gable, Electrical=SBrkr, GarageType=Attchd, SaleCondition=Normal}	=>	{LandContour=Lvl}	0.35123	0.93256	0.37663	1.03385	484
[173]	{LandContour=Lvl, RoofStyle=Gable, Electrical=SBrkr, GarageType=Attchd}	=>	{SaleCondition=Normal}	0.35123	0.82453	0.42598	1.00018	484
[174]	{LandContour=Lvl, Electrical=SBrkr, GarageType=Attchd, SaleCondition=Normal}	=>	{RoofStyle=Gable}	0.35123	0.76704	0.45791	0.98875	484

Algunas reglas interesantes son:

```
IF {LotShape=Reg} THEN {LandCountour=Lvl} con soporte=0.583 y confidence=0.935
IF {LotShape=Reg} THEN {Electrical=SBrkr} con soporte=0.562 y confidence=0.901
IF {GarageType=Attchd} THEN {LandCountour=Lvl} con soporte=0.578 y confidence=0.916
IF {LandCountour=Lvl} THEN {Electrical=SBrkr} con soporte=0.836 y confidence=0.927
IF {LandCountour=Lvl, Electrical=SBrkr, RoofStyle=Gable} THEN {SaleCondition=Normal} con
soporte=0.539 y confidence=0.83
```

Estas reglas, entre otras, tuvieron el mayor nivel de confianza y soporte. Es interesante notar que unas propiedades que parecen no tener relación, dependen una de la otra para suceder.

## Discusión y conclusiones

El data set presentaba 79 variables que contenían información de casas con el propósito de predecir mejor el precio de venta. De estas 79 variables, la mayoría eran de tipo cualitativa. Algunos ejemplos de gráficos de estas variables cualitativas mostraban la proporción o cantidad de datos para esa variable. Por ejemplo, el mes de venta indica que los meses con mayor cantidad de casas vendidas fueron junio y julio. Para las variables cualitativas se utilizaron gráficos de barra, de pie, de proporción y tablas de frecuencia. Por otro lado, en las variables cuantitativas, se utilizaron histogramas, diagrama de caja y bigotes, gráfico de dispersión y gráfico de correlación. Un resultado interesante, por ejemplo, es que GrLivArea y SalePrice presentan una relación lineal. Es importante notar que existen diferentes gráficos para las variables cuantitativas y las cualitativas.

En el análisis de componentes principales, primero se debe analizar si es posible utilizar la técnica de análisis factorial para formar las combinaciones lineales. Para ello, se debe obtener el KMO y el test de Bartlett. En este caso ambos resultados fueron elevados, 0.78 y 6359.1 respectivamente, por lo que sí valía la pena aplicar los componentes principales. Al obtener la matriz de correlación, se notó que las variables que afectan fuertemente al SalePrice son GrLivArea, GarageArea, X1stFlrSF y TotalBsmtSF. Luego, al aplicar los componentes principales, los resultados en sí no son claros. Se tuvieron que normalizar estos datos y obtener la importancia de cada componente principal. Además, se realizaron dos gráficos para determinar la cantidad de componentes que puedan explicar la mayor variabilidad.

Por último, para las reglas de asociación se tomaron en cuenta las variables que representaban una categoría definida. Se utilizó un nivel de confianza de 0.7 y un nivel mínimo de soporte de 0.2. Aunque se obtuvieron bastantes reglas de asociación, las más interesantes fueron las que tuvieron el mayor nivel de confianza y soporte. Como se ha mencionado, al ver las reglas en general, se observó que aunque unas propiedades parecieran no tener relación, dependen una de la otra para suceder.

Se concluye finalmente que a partir de este análisis exploratorio, se deben limpiar los datos, dejar los más importantes, y eliminar correlaciones innecesarias. Existen varias columnas numéricas que contienen muchos NA representados como 0, por lo que hay que tenerlo en consideración al crear una relación numérica con esas columnas. También se debe eliminar la columna de Alley, ya que esta tiene como valor NA para casi todas las filas por lo que no es útil para el análisis.