

Reglas de asociación (Basket Market Analysis)

Transformamos las variables numéricas en categóricas aplicando la función `discretizeDF`.

Cabe resaltar que ahora la base de datos que se utilizará es “dcat” con las variables numéricas como categóricas.

Seguidamente, se transformará “dcat” en un data transactions para poder aplicar el Basket Market Analysis.

```
transactions in sparse format with
5000 transactions (rows) and
113 items (columns)
```

Con el siguiente summary, se puede ver con más detalle lo que se tiene:

```
transactions as itemMatrix in sparse format with
5000 rows (elements/itemsets/transactions) and
113 columns (items) and a density of 0.1415929
```

most frequent items:

```
NAME_EDUCATION_TYPE=Secondary / secondary special
3746
REGION_RATING_CLIENT=2
3641
CODE_GENDER=F
3098
NAME_FAMILY_STATUS=Married
3095
TARGET=0
2865
(Other)
63555
```

```
element (itemset/transaction) length distribution:
sizes
16
5000
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
16	16	16	16	16	16

includes extended item information - examples:

	labels	variables	levels
1	CODE_GENDER=F	CODE_GENDER	F
2	CODE_GENDER=M	CODE_GENDER	M
3	NAME_INCOME_TYPE=Businessman	NAME_INCOME_TYPE	Businessman

includes extended transaction information - examples:

transactionID
1
2
3

Apriori

El primer paso consiste en especificar los parámetros:

El siguiente paso es crear las reglas de asociación:

Apriori

Parameter specification:

```
confidence minval smax arem  aval originalSupport maxtime support minlen
          0.8    0.1    1 none FALSE                TRUE         5   0.002    1
maxlen target  ext
          10  rules TRUE
```

Algorithmic control:

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

Absolute minimum support count: 10

```
set item appearances ...[0 item(s)] done [0.00s].
set transactions ...[110 item(s), 5000 transaction(s)] done [0.00s].
sorting and recoding items ... [98 item(s)] done [0.00s].
creating transaction tree ... done [0.00s].
checking subsets of size 1 2 3 4 5 6 7 8 9 10 done [1.76s].
writing ... [2499418 rule(s)] done [0.55s].
creating S4 object ... done [1.13s].
```

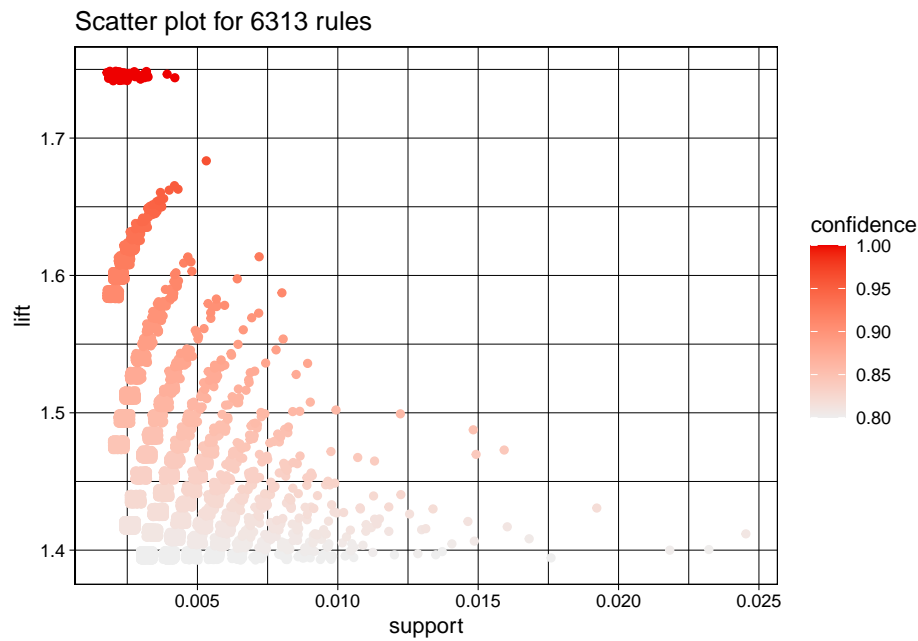
Dividimos las reglas de asociación obtenidas según lo consecuente que es la variable respuesta. La variable respuesta es **TARGET**, que toma valores de 1 o 0.

Se eliminan las reglas redundantes en ambos casos:

```
Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.00000 0.01065 0.02847 0.02959 0.05815 0.06071
```

```
Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
2.000e-09 4.957e-04 1.916e-03 2.117e-03 3.702e-03 4.951e-03
```

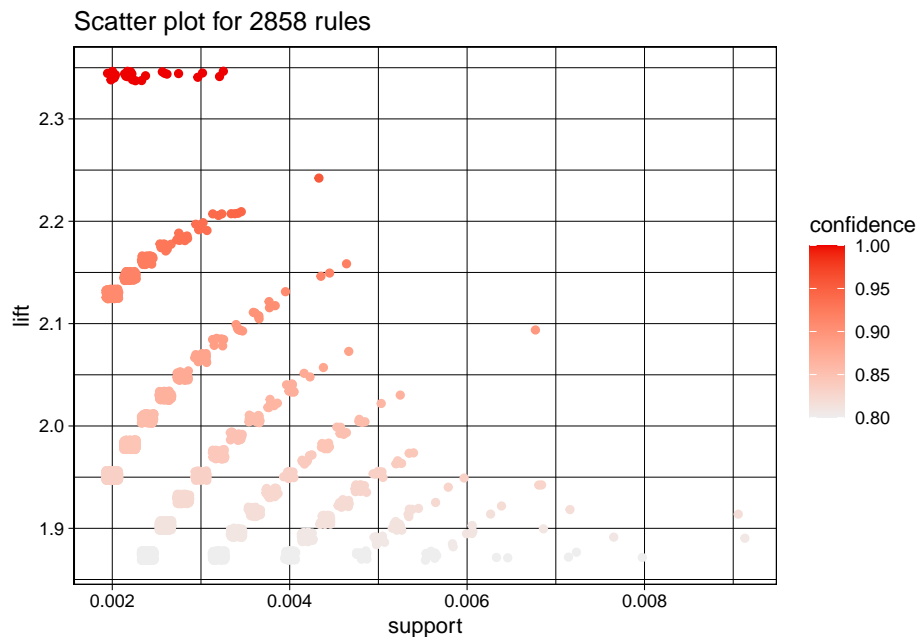
Figura 52: Scatter plot for 6313 rules



Como se puede ver, el primer gráfico muestra la matriz de puntos de las reglas de asociación filtrada respecto la métrica lift. Las reglas de asociación de interés corresponden a los puntos con un color rojo de mayor intensidad (confianza que supere la mínima, 0.8) y se aprecia, estas reglas se sitúan en el gráfico con un soporte mayor al mínimo (0.002).

En el último gráfico se ve algo parecido, aquí las reglas de asociación que interesan corresponden a los puntos con una intensidad roja mayor y los puntos más grandes, que correspondieran a las reglas que tienen un soporte superior al mínimo (0.002).

Figura 53: Scatter plot for 2858 rules



Estos gráficos se interpretan de manera igual a los anteriores vistos.

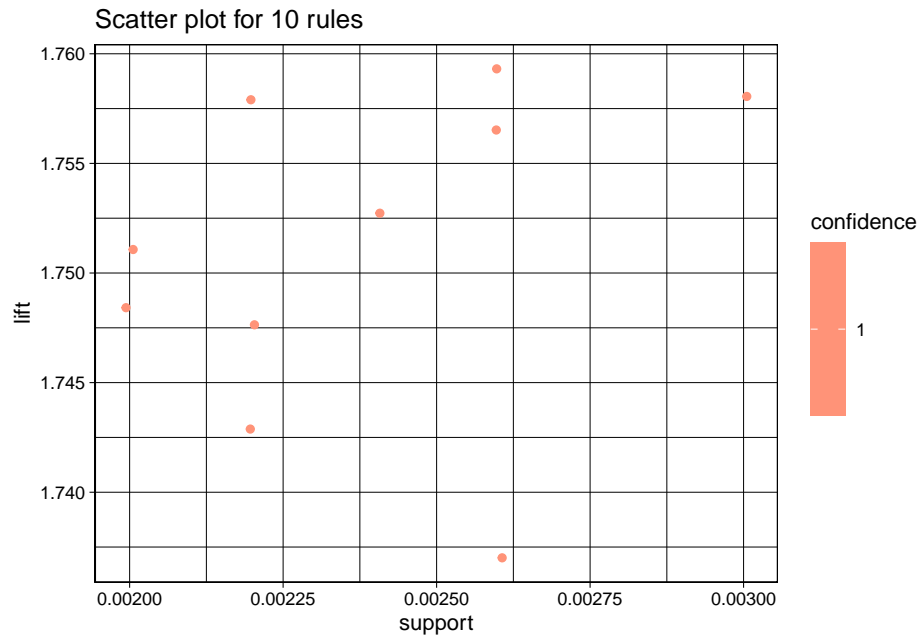
Con Target = 0 se obtienen 6313 reglas y con Tagret = 1 2858 reglas. Con la gran cantidad de reglas, la atención se centra en las 10 primeras reglas en cada caso con mayor lift.

Por tanto, se ven las 10 primeras reglas en cada caso con mayor lift, es decir, van ordenadas de forma decreciente siendo la primera la que tiene una mayor asociación encontrada con la variable respuesta, y se grafican en cada caso.

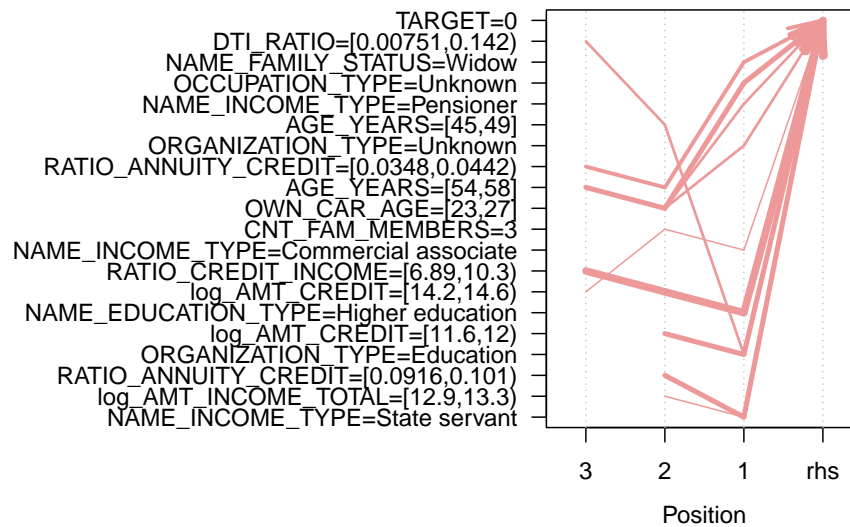
	lhs	rhs	support	confidence	coverage	lift	count
[1]	{NAME_INCOME_TYPE=State servant, log_AMT_INCOME_TOTAL=[12.9,13.3]}	=> {TARGET=0}	0.0020	1	0.0020	1.745201	10
[2]	{NAME_INCOME_TYPE=State servant, RATIO_ANNUITY_CREDIT=[0.0916,0.101]}	=> {TARGET=0}	0.0026	1	0.0026	1.745201	13
[3]	{ORGANIZATION_TYPE=Education, log_AMT_CREDIT=[11.6,12]}	=> {TARGET=0}	0.0026	1	0.0026	1.745201	13
[4]	{NAME_EDUCATION_TYPE=Higher education, log_AMT_CREDIT=[14.2,14.6], RATIO_CREDIT_INCOME=[6.89,10.3]}	=> {TARGET=0}	0.0030	1	0.0030	1.745201	15
[5]	{NAME_INCOME_TYPE=Commercial associate, CNT_FAM_MEMBERS=3, log_AMT_CREDIT=[14.2,14.6]}	=> {TARGET=0}	0.0020	1	0.0020	1.745201	10
[6]	{NAME_FAMILY_STATUS=Widow, AGE_YEARS=[54,58], RATIO_ANNUITY_CREDIT=[0.0348,0.0442]}	=> {TARGET=0}	0.0024	1	0.0024	1.745201	12
[7]	{ORGANIZATION_TYPE=Unknown, OWN_CAR_AGE=[23,27], AGE_YEARS=[54,58]}	=> {TARGET=0}	0.0022	1	0.0022	1.745201	11
[8]	{NAME_INCOME_TYPE=Pensioner,						

	OWN_CAR_AGE=[23,27], AGE_YEARS=[54,58]}	=> {TARGET=0}	0.0022	1	0.0022	1.745201	11
[9]	{OCCUPATION_TYPE=Unknown, OWN_CAR_AGE=[23,27], AGE_YEARS=[54,58]}	=> {TARGET=0}	0.0026	1	0.0026	1.745201	13
[10]	{ORGANIZATION_TYPE=Education, AGE_YEARS=[45,49], DTI_RATIO=[0.00751,0.142)}	=> {TARGET=0}	0.0022	1	0.0022	1.745201	11

Figura 54: Scatter plot for 10 rules

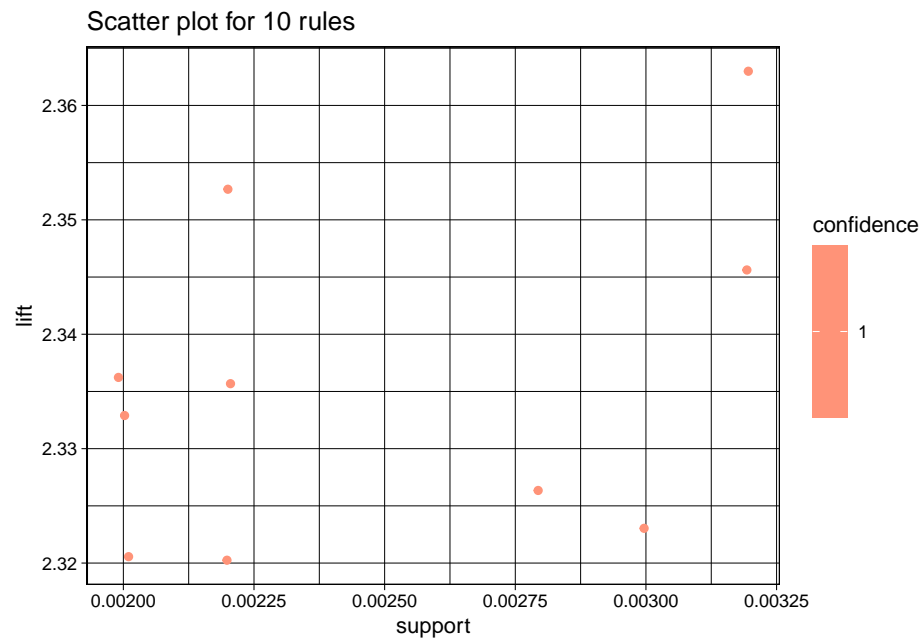


Parallel coordinates plot for 10 rules

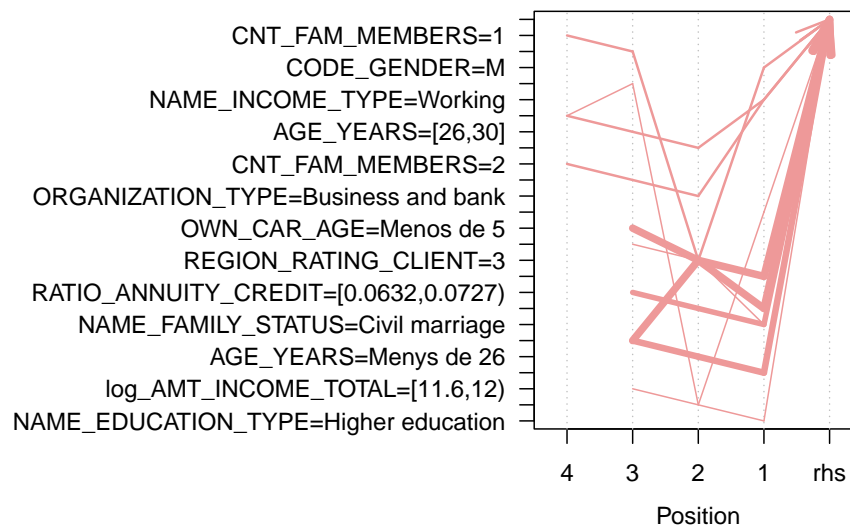


	lhs	rhs	support	confidence	coverage	lift	count
[1]	{NAME_EDUCATION_TYPE=Higher education, OWN_CAR_AGE=[23,27], log_AMT_INCOME_TOTAL=[11.6,12]}	=> {TARGET=1}	0.0020	1	0.0020	2.34192	10
[2]	{log_AMT_CREDIT=[12.4,12.9), AGE_YEARS=Menys de 26, RATIO_ANNUITY_CREDIT=[0.0727,0.0821)}	=> {TARGET=1}	0.0030	1	0.0030	2.34192	15
[3]	{NAME_FAMILY_STATUS=Civil marriage, OCCUPATION_TYPE=Low-mid skill laborers, RATIO_ANNUITY_CREDIT=[0.0632,0.0727)}	=> {TARGET=1}	0.0028	1	0.0028	2.34192	14
[4]	{OCCUPATION_TYPE=Low skill laborers, REGION_RATING_CLIENT=3, RATIO_ANNUITY_CREDIT=[0.0727,0.0821)}	=> {TARGET=1}	0.0032	1	0.0032	2.34192	16
[5]	{NAME_FAMILY_STATUS=Civil marriage, REGION_RATING_CLIENT=3, AGE_YEARS=[45,49]}	=> {TARGET=1}	0.0020	1	0.0020	2.34192	10
[6]	{OCCUPATION_TYPE=Low-mid skill laborers, REGION_RATING_CLIENT=3, OWN_CAR_AGE=Menos de 5}	=> {TARGET=1}	0.0032	1	0.0032	2.34192	16
[7]	{NAME_INCOME_TYPE=Working, ORGANIZATION_TYPE=Business and bank, OWN_CAR_AGE=[28,32], CNT_FAM_MEMBERS=2}	=> {TARGET=1}	0.0022	1	0.0022	2.34192	11
[8]	{NAME_INCOME_TYPE=Working, ORGANIZATION_TYPE=Trade and telecom, AGE_YEARS=[26,30], DTI_RATIO=[0.142,0.276)}	=> {TARGET=1}	0.0022	1	0.0022	2.34192	11
[9]	{OCCUPATION_TYPE=Unknown, OWN_CAR_AGE=[23,27], log_AMT_INCOME_TOTAL=[11.2,11.6), DTI_RATIO=[0.142,0.276)}	=> {TARGET=1}	0.0020	1	0.0020	2.34192	10
[10]	{CODE_GENDER=M, REGION_RATING_CLIENT=3, OWN_CAR_AGE=[19,22], CNT_FAM_MEMBERS=1}	=> {TARGET=1}	0.0022	1	0.0022	2.34192	11

Figura 55: Scatter plot for 10 rules



Parallel coordinates plot for 10 rules



ECLAT

Para este apartado, se crearán las reglas de asociación con ECLAT.

Eclat

parameter specification:

```

tidLists support minlen maxlen          target  ext
  FALSE   0.002      1      10 frequent itemsets TRUE

```

```

algorithmic control:
  sparse sort verbose
      7   -2    TRUE

```

Absolute minimum support count: 10

```

create itemset ...
set transactions ... [110 item(s), 5000 transaction(s)] done [0.00s].
sorting and recoding items ... [98 item(s)] done [0.00s].
creating bit matrix ... [98 row(s), 5000 column(s)] done [0.00s].
writing ... [1992604 set(s)] done [0.66s].
Creating S4 object ... done [0.55s].

```

set of 51080 rules

```

rule length distribution (lhs + rhs):sizes
  2    3    4    5    6    7    8    9    10
  3  111 1756 8211 15442 14643 7758 2596 560

```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
2.00	6.00	7.00	6.55	7.00	10.00

summary of quality measures:

support	confidence	lift	itemset
Min. :0.002000	Min. :0.8000	Min. :1.396	Min. : 2
1st Qu.:0.002200	1st Qu.:0.8235	1st Qu.:1.437	1st Qu.: 271311
Median :0.002400	Median :0.8462	Median :1.477	Median : 670530
Mean :0.002764	Mean :0.8626	Mean :1.505	Mean : 767225
3rd Qu.:0.003000	3rd Qu.:0.9091	3rd Qu.:1.587	3rd Qu.:1289119
Max. :0.024600	Max. :1.0000	Max. :1.745	Max. :1987637

mining info:

```

data ntransactions support
  tr           5000   0.002

```

call

```

eclat(data = tr, parameter = list(support = soporte_minimo, minlen = 1, maxlen = tamanyo_conjunto))
confidence
  0.8

```

set of 10520 rules

```

rule length distribution (lhs + rhs):sizes
  3    4    5    6    7    8    9    10
  4  206 1324 3155 3432 1864 479 56

```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
3.000	6.000	7.000	6.672	7.000	10.000

summary of quality measures:

support	confidence	lift	itemset
Min. :0.002000	Min. :0.8000	Min. :1.874	Min. : 694
1st Qu.:0.002000	1st Qu.:0.8235	1st Qu.:1.929	1st Qu.: 418570
Median :0.002400	Median :0.8462	Median :1.982	Median : 917806
Mean :0.002532	Mean :0.8607	Mean :2.016	Mean : 882066
3rd Qu.:0.002800	3rd Qu.:0.9091	3rd Qu.:2.129	3rd Qu.:1238898
Max. :0.009200	Max. :1.0000	Max. :2.342	Max. :1976927

mining info:

```
data ntransactions support
tr      5000    0.002
```

```
call
eclat(data = tr, parameter = list(support = soporte_minimo, minlen = 1, maxlen = tamanyo_conjunto))
confidence
0.8
```

	lhs	rhs	support	confidence	lift	itemset
[1]	{NAME_INCOME_TYPE=State servant, log_AMT_INCOME_TOTAL=[12.9,13.3]}	=> {TARGET=0}	0.0020	1	1.745201	10346
[2]	{NAME_EDUCATION_TYPE=Higher education, log_AMT_CREDIT=[14.2,14.6), RATIO_CREDIT_INCOME=[6.89,10.3]}	=> {TARGET=0}	0.0030	1	1.745201	13160
[3]	{NAME_INCOME_TYPE=Commercial associate, CNT_FAM_MEMBERS=3, log_AMT_CREDIT=[14.2,14.6)}	=> {TARGET=0}	0.0020	1	1.745201	14725
[4]	{NAME_INCOME_TYPE=Commercial associate, NAME_FAMILY_STATUS=Married, ORGANIZATION_TYPE=Personal services, RATIO_CREDIT_INCOME=[0.125,3.51]}	=> {TARGET=0}	0.0030	1	1.745201	23611
[5]	{CODE_GENDER=F, OCCUPATION_TYPE=Low-mid skill laborers, REGION_RATING_CLIENT=2, RATIO_ANNUITY_CREDIT=[0.101,0.111)}	=> {TARGET=0}	0.0028	1	1.745201	62972
[6]	{NAME_INCOME_TYPE=State servant, RATIO_ANNUITY_CREDIT=[0.0916,0.101)}	=> {TARGET=0}	0.0026	1	1.745201	66662
[7]	{NAME_FAMILY_STATUS=Widow, AGE_YEARS=[54,58], RATIO_ANNUITY_CREDIT=[0.0348,0.0442]}	=> {TARGET=0}	0.0024	1	1.745201	95632
[8]	{ORGANIZATION_TYPE=Education, log_AMT_CREDIT=[11.6,12)}	=> {TARGET=0}	0.0026	1	1.745201	118213
[9]	{NAME_INCOME_TYPE=Pensioner, OWN_CAR_AGE=[23,27], AGE_YEARS=[54,58]}	=> {TARGET=0}	0.0022	1	1.745201	132450
[10]	{ORGANIZATION_TYPE=Unknown, OWN_CAR_AGE=[23,27], AGE_YEARS=[54,58]}	=> {TARGET=0}	0.0022	1	1.745201	132466

	lhs	rhs	support	confidence	lift	itemset
[1]	{NAME_INCOME_TYPE=Working, ORGANIZATION_TYPE=Business and bank,					

	OWN_CAR_AGE=[28,32], CNT_FAM_MEMBERS=2}	=> {TARGET=1}	0.0022	1	2.34192	20786
[2]	{NAME_INCOME_TYPE=Working, ORGANIZATION_TYPE=Trade and telecom, AGE_YEARS=[26,30], DTI_RATIO=[0.142,0.276)}	=> {TARGET=1}	0.0022	1	2.34192	78683
[3]	{OCCUPATION_TYPE=Unknown, OWN_CAR_AGE=[23,27], log_AMT_INCOME_TOTAL=[11.2,11.6), DTI_RATIO=[0.142,0.276)}	=> {TARGET=1}	0.0020	1	2.34192	135824
[4]	{NAME_EDUCATION_TYPE=Higher education, OWN_CAR_AGE=[23,27], log_AMT_INCOME_TOTAL=[11.6,12)}	=> {TARGET=1}	0.0020	1	2.34192	137658
[5]	{CODE_GENDER=M, REGION_RATING_CLIENT=3, OWN_CAR_AGE=[19,22], CNT_FAM_MEMBERS=1}	=> {TARGET=1}	0.0022	1	2.34192	204659
[6]	{log_AMT_CREDIT=[12.4,12.9), AGE_YEARS=Menys de 26, RATIO_ANNUITY_CREDIT=[0.0727,0.0821)}	=> {TARGET=1}	0.0030	1	2.34192	213819
[7]	{NAME_FAMILY_STATUS=Single / not married, OCCUPATION_TYPE=Low skill laborers, log_AMT_INCOME_TOTAL=[11.6,12), AGE_YEARS=Menys de 26}	=> {TARGET=1}	0.0020	1	2.34192	217788
[8]	{NAME_FAMILY_STATUS=Separated, OCCUPATION_TYPE=Low-mid skill laborers, log_AMT_INCOME_TOTAL=[12,12.5), RATIO_CREDIT_INCOME=[0.125,3.51)}	=> {TARGET=1}	0.0024	1	2.34192	244609
[9]	{NAME_FAMILY_STATUS=Civil marriage, OCCUPATION_TYPE=Low-mid skill laborers, RATIO_ANNUITY_CREDIT=[0.0632,0.0727)}	=> {TARGET=1}	0.0028	1	2.34192	303233
[10]	{OCCUPATION_TYPE=Low-mid skill laborers, CNT_FAM_MEMBERS=2, RATIO_ANNUITY_CREDIT=[0.0632,0.0727), DTI_RATIO=[0.00751,0.142)}	=> {TARGET=1}	0.0026	1	2.34192	319187