

INSTITUTO TECNOLÓGICO DE CULIACÁN

Unidad 3:

Minería de datos

MATERIA:

Temas Selectos de Base de Datos

ACTIVIDAD:

Practica 1 Comparativa de algoritmos de Mineria

CARRERA:

Ingeniería en Sistemas Computacionales

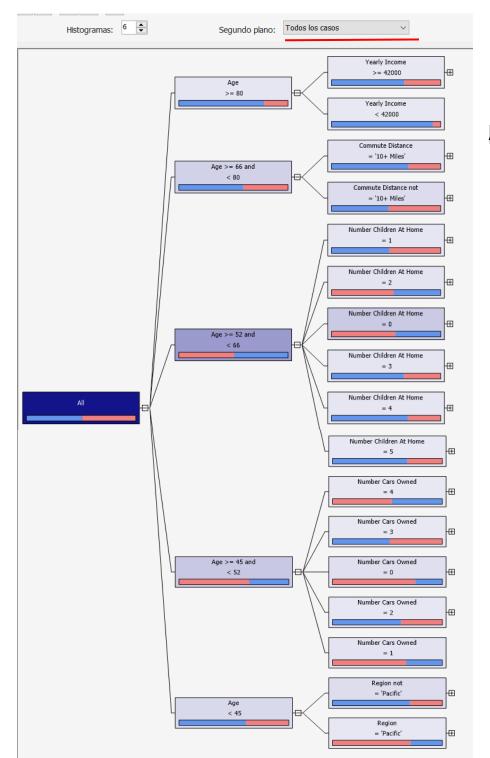
MAESTRO:

MC.Daniel Esparza Soto

INTEGRANTES EQUIPO:

Castro Cruz Luis Daniel Felix Fierro Ryan Guadalupe

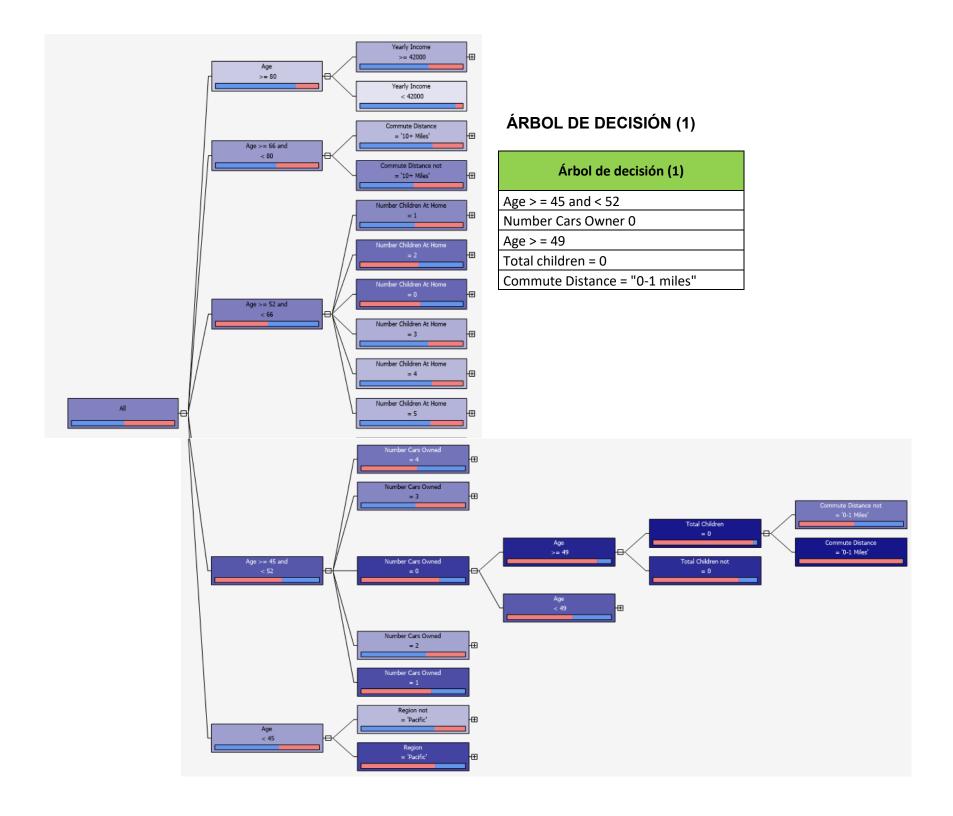
Flores Mascareño America Citlaly

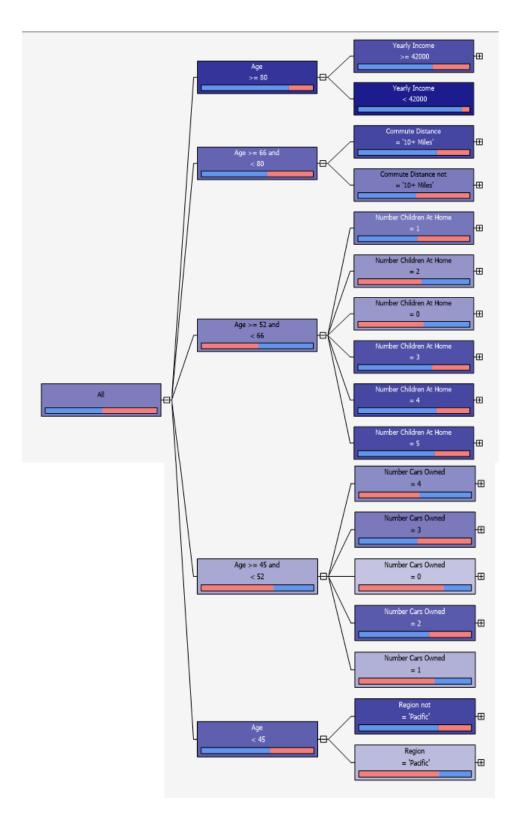


ÁRBOL DE DECISIÓN

(Todos los casos)

	Árbol de decisión	
Arbol de decisión (1)	Arbol de decisión (0)	Arbol de decisión/RedDependencias
Age > = 45 and < 52	Age >= 80	Age
Number Cars Owner 0	Yearly income < 42000	Region
Age > = 49		Number Children At Home
Total children = 0		Number Cars Owned
Commute Distance = "0-1 miles"		Yearly income
		Marital Status
		Commute Distance
		English Ocupation
		English Education
		English Education
		House owner Flag





ÁRBOL DE DECISIÓN (0)

Arbol de decisión (0)

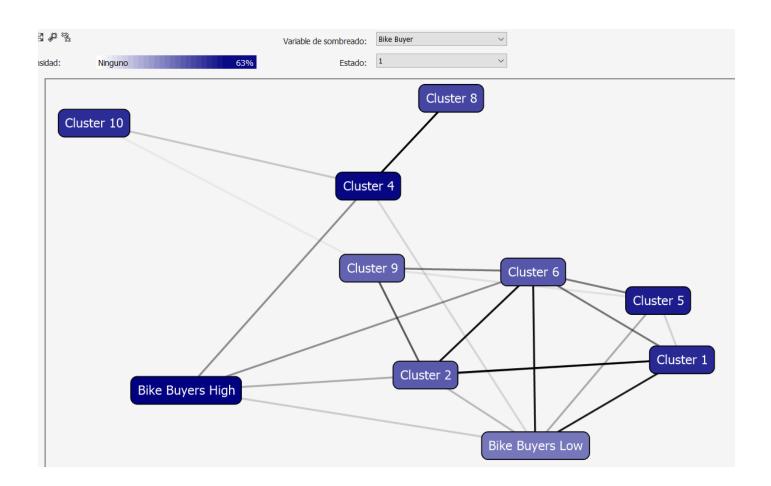
Age >= 80

Yearly income < 42000

Arbol de decisión/RedDependencias ÁRBOL DE DECISIÓN / RED DEPENDENCIAS Age Region Number Children At Home **Number Cars Owned** Yearly income **Marital Status Commute Distance English Ocupation** Region **English Education** Number Cars Owned **English Education** House owner Flag Yearly Income House Owner Flag Commute Distance Bike Buyer **English Education English Occupation** Marital Status Age Total Children Number Children At Home

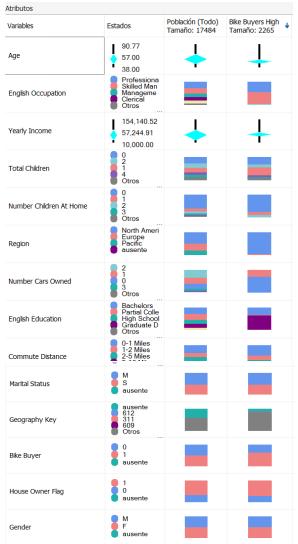
DIAGRAMA DE CLÚSTER

		Cluster	
Perfil de cluster: Bike Buyer High	Perfil de cluster: Bike Buyer Low	Caracteristicas de cluster: Bike Buyer High	Caracteristicas de cluster: Bike Buyer Low
Age	Region	Region North America	Region: North America
English Ocupation	Age	Number Children At Home: 0	Age: 38 - 49
Yearly Income	English Education	House owner Flag: 1	Number Children At Home: 0
Total children	Number Cars Owner	Number Cars Owner: 0	Total children: 0
Number Children At		English Education: Graduate	
Home	Yearly Income	Degree	Number Cars Owner: 2
Region	English Ocupation	Bike Buyer: 1	English Ocupation: Skilled Manual
Number Cars Owned	Commute Distance	Age: 50 -56	Bike Buyer: 0
English education	Number Children At Home	Marital Status: M	Marital Status: S
Commute Distance	Bike Buyer	English Ocupation: Skilled Manual	Yearly income: 35,459.9 - 57,244.9
Marital Status	Total Children	Yearly income: 57,244.9 - 79,029.9	Commute Distance: 5-10 miles
Geography key	Geography Key	Gender: F	House owner Flag: 1
Bike Buyer	House owner Flag	Gender: M	English Education: Partial College
House Owner Flag	Marital Status	Marital Status: S	Gender: M
Gender	Gender		Gender: F



Perfiles de Cluster

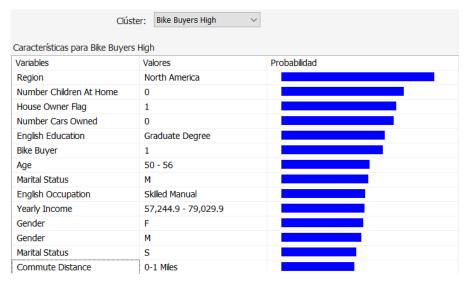
Perfil de cluster: Bike Buyer High	Perfil de cluster: Bike Buyer Low			
Age	Region			
English Ocupation	Age			
Yearly Income	English Education			
Total children	Number Cars Owner			
Number Children At Home	Yearly Income			
Region	English Ocupation			
Number Cars Owned	Commute Distance			
English education	Number Children At Home			
Commute Distance	Bike Buyer			
Marital Status	Total Children			
Geography key	Geography Key			
Bike Buyer	House owner Flag			
House Owner Flag	Marital Status			
Gender	Gender			

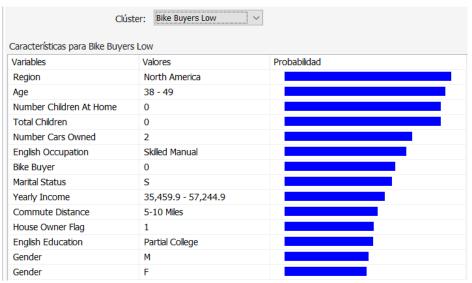




Características de Cluster

Caracteristicas de cluster: Bike Buyer High	Caracteristicas de cluster: Bike Buyer Low
Region North America	Region: North America
Number Children At Home: 0	Age: 38 - 49
House owner Flag: 1	Number Children At Home: 0
Number Cars Owner: 0	Total children: 0
English Education: Graduate Degree	Number Cars Owner: 2
Bike Buyer: 1	English Ocupation: Skilled Manual
Age: 50 -56	Bike Buyer: 0
Marital Status: M	Marital Status: S
English Ocupation: Skilled Manual	Yearly income: 35,459.9 - 57,244.9
Yearly income: 57,244.9 - 79,029.9	Commute Distance: 5-10 miles
Gender: F	House owner Flag: 1
Gender: M	English Education: Partial College
Marital Status: S	Gender: M
	Gender: F





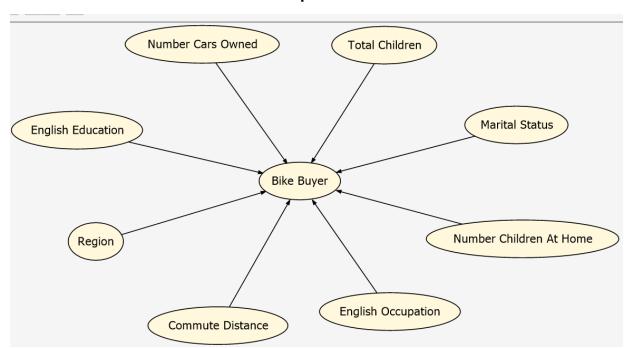
Ficha distinción del clúster

Variables	Valores	Favorece Bike Buyer	Favorece Bike Buyer
Number Cars Owned	2	•	,
Age	38 - 45		
Age	46 - 108		
Number Cars Owned	0		
English Education	Graduate Degree		
Commute Distance	5-10 Miles		
Total Children	0		
English Education	Partial College		
English Education	High School		
Total Children	1		
Commute Distance	2-5 Miles		
English Education	Bachelors		
Yearly Income	46,922.6 - 170,000.0		
Yearly Income	10,000.0 - 46,922.6		
English Occupation	Professional		
English Occupation	Clerical		
Commute Distance	0-1 Miles		
Total Children	2		
Number Children At Home	1		
Bike Buyer	0		
Bike Buyer	1	_	
English Education	Partial High School		
Number Children At Home	2	_	
Number Children At Home	0		
English Occupation	Skilled Manual		•
Number Children At Home	3		I
Total Children	4		

Modelo Bayes Naive

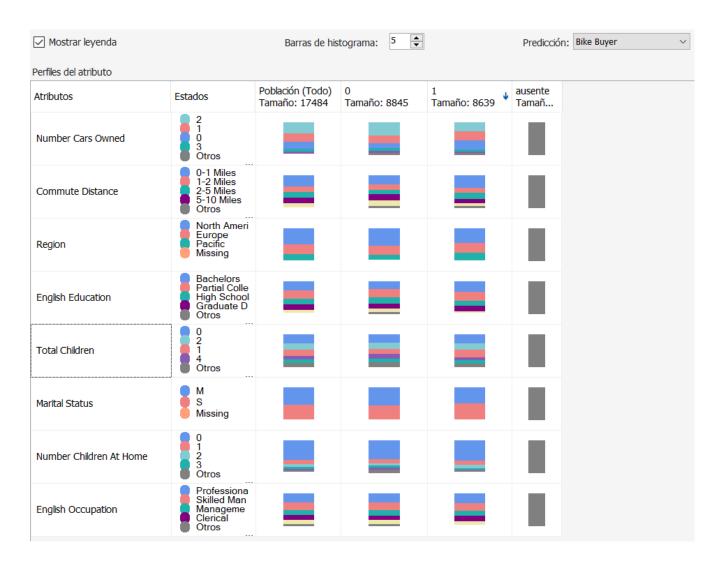
	Bayes Naive				
Red de dependencias	Perfil de atributo	Caracteristica del atributo: si compraron			
Number Cars Owner	Number Cars Owner	Number Children At Home 0			
Commute Distance	Commute Distance	Marital Status M			
Total children	Region	Marital Status S			
English Education	English Education	Region NA			
Number Children At Home	Total children	Commute Distance 0-1 miles			
Region	Marital Status	English Ocupation Bacherol			
Marital Status	Number Children At Home	English Ocupation Professional			
English Ocupation	English Ocupation	Region Europe			
		Number Cars Owner 0			
		Number Cars Owner 1			
		Total children 0			
		Number Cars Owner 2			
		English Education Partial College			

Red de Dependencias



Perfil de atributo

Perfil de atributo
Number Cars Owner
Commute Distance
Region
English Education
Total children
Marital Status
Number Children At Home
English Ocupation

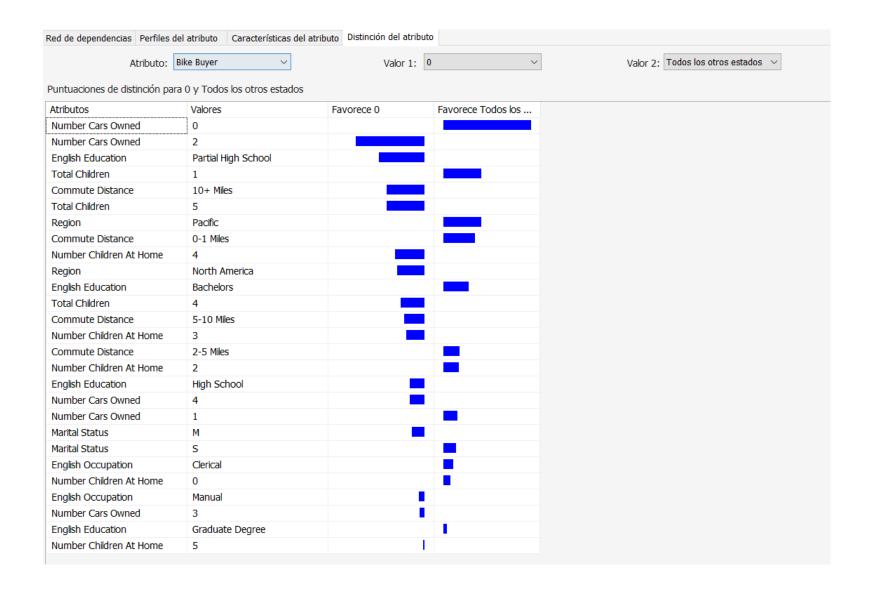


Características del atributo: Sí compraron

Caracteristica del atributo: si compraron
Number Children At Home: 0
Marital Status: M
Marital Status: S
Region: North America
Commute Distance: 0-1 miles
English Ocupation: Bachelors
English Ocupation: Professional
Region: Europe
Number Cars Owner: 0
Number Cars Owner: 1
Total children: 0
Number Cars Owner: 2
English Education: Partial College

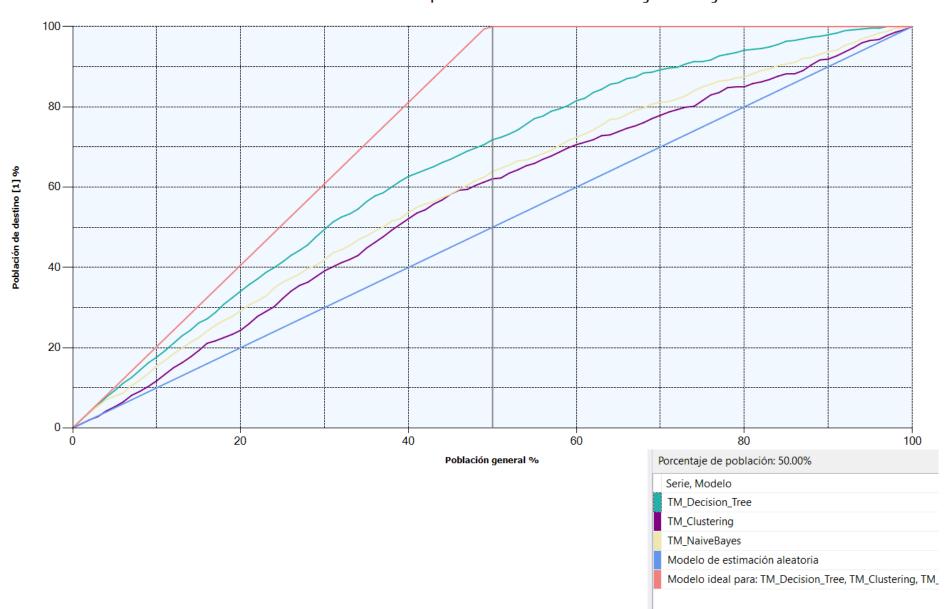


Distinción del atributo



GRAFICA DE ELEVACIÓN

Gráfico de elevación de minería de datos para estructura de minería de datos: Targeted Mailing

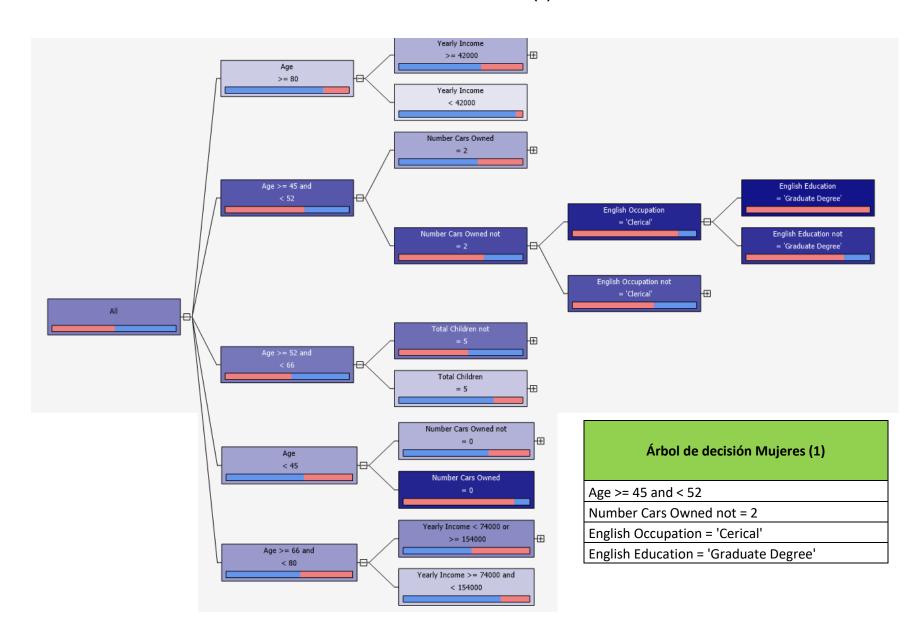


PROBANDO UN MODELO DE FILTRADO

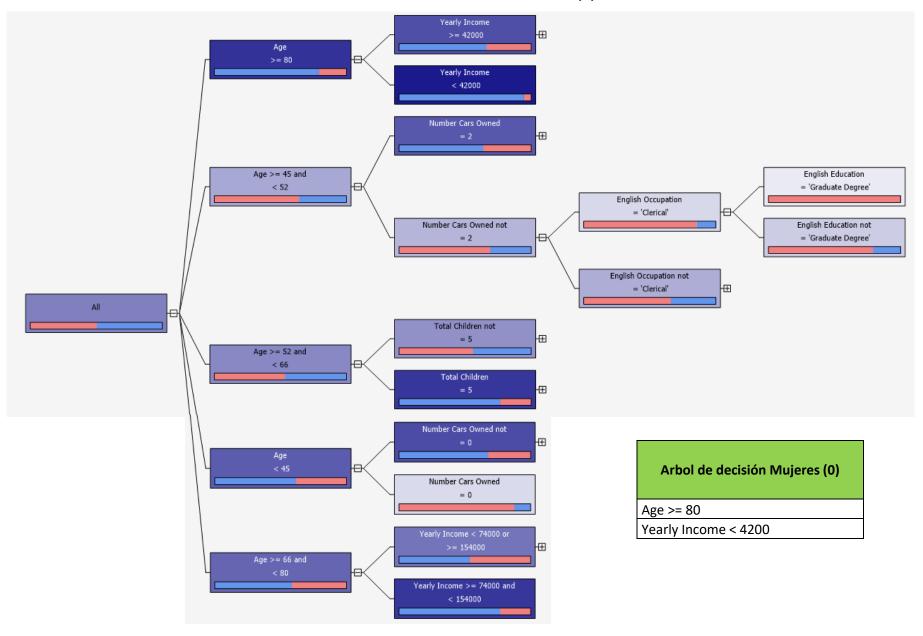
Árbol de decisión Male/Female

	Arbol de decisión				
Arbol de decisión Hombres (0)	Arbol de decisión Hombres (1)	Arbol de decisión Mujeres (0)	Arbol de decisión Mujeres (1)	Arbol de decisión/RedDependencias Hombres	Arbol de decisión/RedDependencias Mujeres
Age >=80	Age >= 45 and < 52	Age >= 80	Age >= 45 and < 52	Age	Age
Yearly Income < 5800	Number Cars Owned not = 2	Yearly Income < 4200	Number Cars Owned not = 2	Region	Numbers Cars Owned
	English Occupation = 'Clerical'		English Occupation = 'Cerical'	Number Children At Home	Total Children
	Number Cars Owned = 0		English Education = 'Graduate Degree'	Yearly Income	Yearly Income
	Total Children = 0			Number Cars Owned	Commute Distance
				English Occupation	Number Children at Home
				Commute Distance	English Occupation
				Marital Status	English Education
				Total Children	Region
				English Education	Marital Status

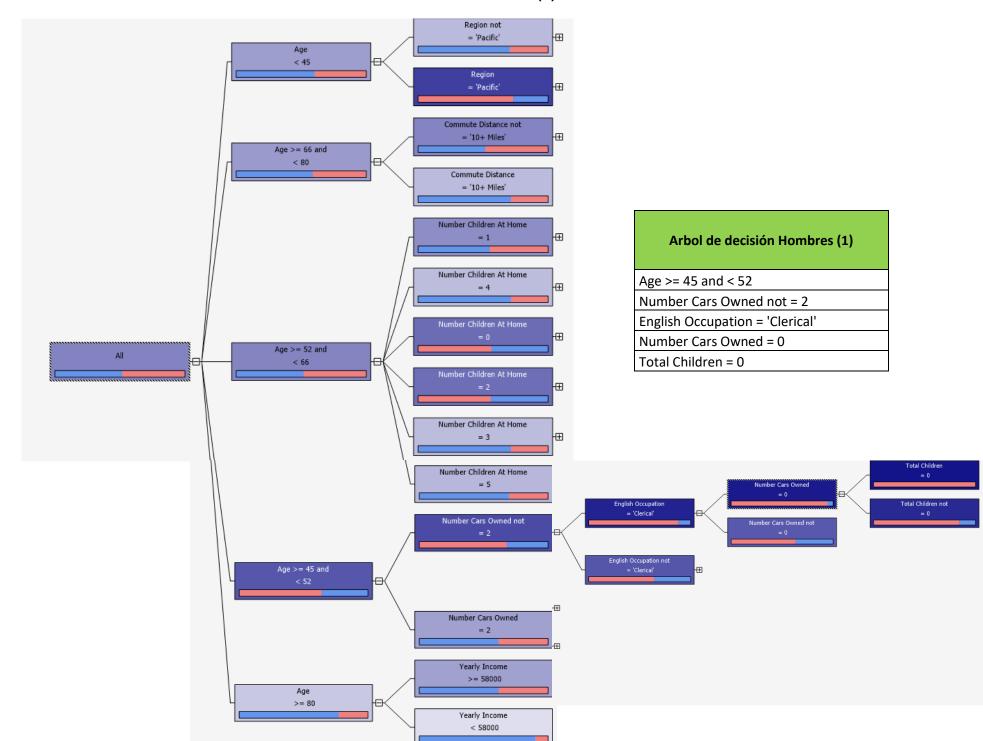
ÁRBOL DE DECISIÓN FEMALE (1)



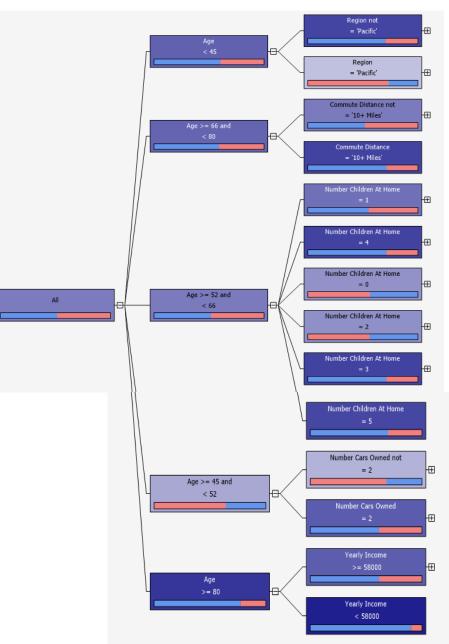
ÁRBOL DE DECISIÓN FEMALE (0)



ÁRBOL DE DECISIÓN MALE (1)



ÁRBOL DE DECISIÓN MALE (0)



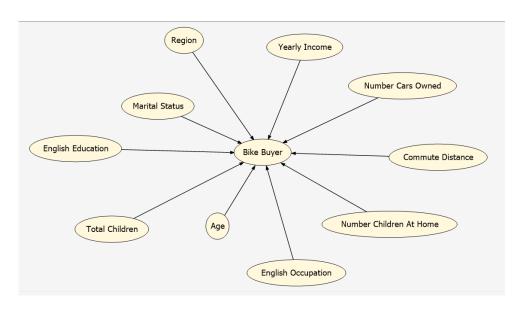
Arbol de decisión Hombres (0)

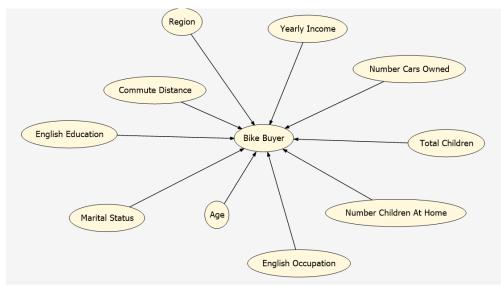
Age >=80

Yearly Income < 5800

DEPENDENCIAS MALE/FEMALE

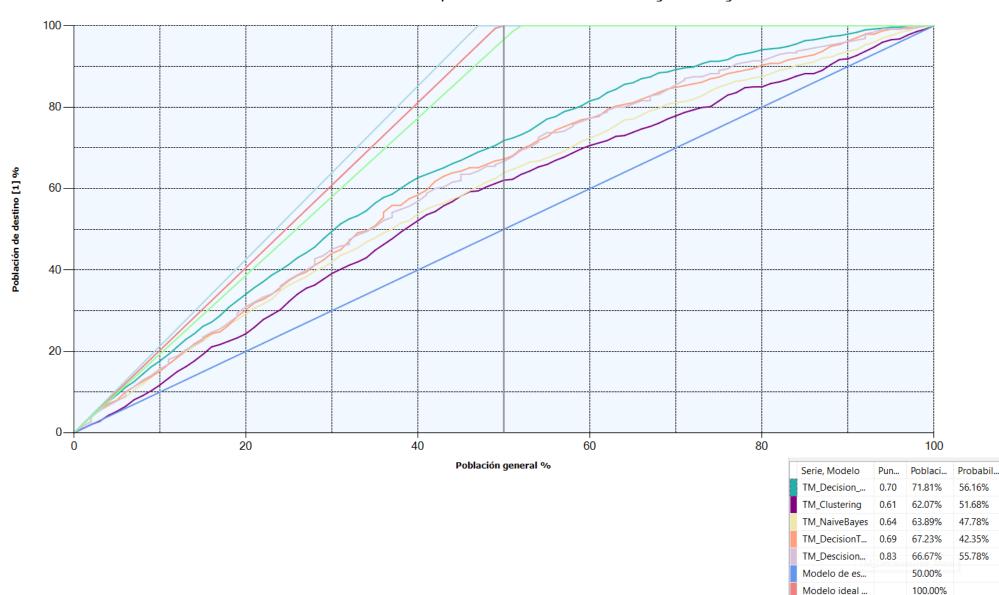
Arbol de decisión/RedDependencias Hombres	Arbol de decisión/RedDependencias Mujeres				
Age	Age				
Region	Numbers Cars Owned				
Number Children At Home	Total Children				
Yearly Income	Yearly Income				
Number Cars Owned	Commute Distance				
English Occupation	Number Children at Home				
Commute Distance	English Occupation				
Marital Status	English Education				
Total Children	Region				
English Education	Marital Status				





GRAFICA DE ELEVACIÓN FEMALE

Gráfico de elevación de minería de datos para estructura de minería de datos: Targeted Mailing



Modelo ideal ...

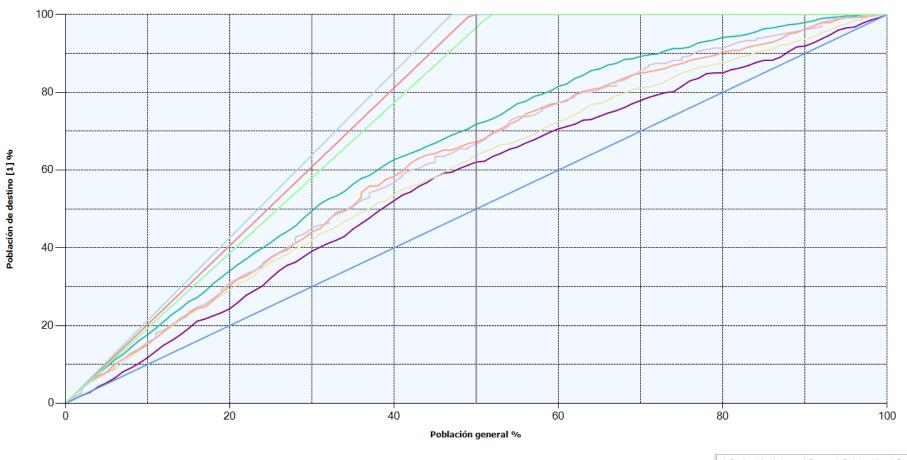
Modelo ideal ...

100.00%

96.67%

GRAFICA DE ELEVACIÓN MALE

Gráfico de elevación de minería de datos para estructura de minería de datos: Targeted Mailing



Pun	Població	Probabili
0.70	71.81%	56.16%
0.61	62.07%	51.68%
0.64	63.89%	47.78%
0.69	67.23%	42.35%
0.83	66.67%	55.78%
	50.00%	
	100.00%	
	100.00%	
	96.67%	
	0.70 0.61 0.64 0.69	0.70 71.81% 0.61 62.07% 0.64 63.89% 0.69 67.23% 0.83 66.67% 50.00% 100.00%

Predicción de árbol de decisión

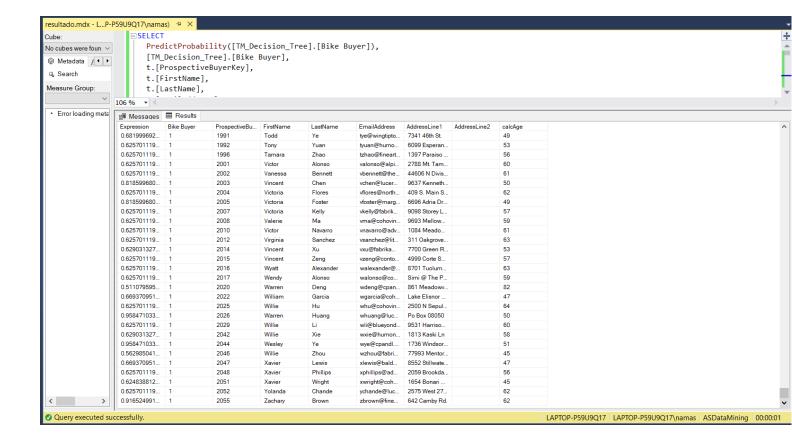
Resultado

7-6		datos 🧪 Modek	os de minería de	e datos 🔏 Vis	or de modelos de i	minería de datos	Gráfico de	precisión de minería de datos	Predicción de modelo de minería de datos
Expression	Bike Buver	Prospective	FirstName	LastName	EmailAddress	AddressLine1	AddressLine2	calcAge	
0.5629850	1	6	Angel	Bell	abell@thep	840 Charlot		45	
0.8185996	1	7	Anna	Bennett	abennett@	312 Via Del		49	
0.6257011	1	9	Arturo	Bhat	abhat@ada	7040 Isabel		53	
0.5629850	1	11	Abigail	Bryant	abryant@n	2639 Anch		46	
0.6257011	1	15	Adam	Carter	acarter@m	4082 Rosly		60	
0.6693709	1	17	Alberto	Castro	acastro@th	2696 Santa		47	
0.6693709	1	18	Alicia	Chande	achande@a	5935 Isabel		51	
0.5524897	1	19	April	Chande	achande@c	7939 Bayvi		53	
0.5629850	1	23	Alexander	Clark	aclark@hu	3243 Lanto		51	
0.6257011	1	29	Alexandria	Cooper	acooper@c	1624 Carlisl		52	
0.6257011	1	30	Alexis	Davis	adavis@co	6900 Willia		60	
0.9165249	1	31	Abigail	Davis	adavis@fab	70 N.w. Plaza		64	
0.8185996	1	32	Alicia	Deng	adeng@cpa	8958 Carlet		49	
0.6693709	1	36	Andre	Engineer	aengineer@	North 9327		49	
0.6693709	1	37	Andre	Fernandez	afernandez	2292 Sprin		50	
0.5110795	1	38	Adriana	Fernandez	afernandez	1314 Green		82	
0.6257011	1	40	Adam	Foster	afoster@wo	4524 Fernd		63	
0.6257011	1	47	Alexandra	Gonzalez	agonzalez@	9075 Calle		59	
0.6693709	1	48	Abigail	Gray	agray@con	78025 E. M		51	
0.8185996	1	50	Alyssa	Griffin	agriffin@co	6916 Azores		49	
0.6248388	1	53	Alex	Hall	ahall@coho	9115 Arthu		46	
0.6243700	1	55	Abigail	Hall	ahall@north	8036 Sum		47	
0.6257011	1	57	Alejandro	He	ahe@fabrik	Pyramid Mall		59	
0.6257011	1	67	Ana	Jenkins	ajenkins@h	4769 Buch	Unit G 202	65	
0.6257011	1	70	Armando	Jimenez	ajimenez@	3880 Batacao		57	
0.5110795	1	71	Anna	Kelly	akelly@adat	7750 E Mcd		87	
0.5524897	1	77	Aimee	Liang	allang@fabri	1601 Crow		54	
0.6693709	1	78	Aimee	Lin	alin@litwarei	1133 Conc		47	
0.6257011	1	79	Amy	Liu	aliu@fabrika	6837 Pirate		56	
0.6693709	1	80	Arianna	Long	along@coh	1349 Palm		50	
0.9584710	1	84	Arthur	Lopez	alopez@co	7549 Long		51	
0.6257011	1	85	Abby	Malhotra	amalhotra	1019 Carlet		53	
0.6248388	1	86	Andre	Malhotra	amalhotra	6968 Mildre		47	

Consulta en análisis services

```
SELECT
  PredictProbability([TM_Decision_Tree].[Bike Buyer]),
  [TM_Decision_Tree].[Bike Buyer],
  t.[ProspectiveBuyerKey],
  t.[FirstName],
  t.[LastName],
  t.[EmailAddress],
  t.[AddressLine1],
  t.[AddressLine2],
  t.[calcAge]
From
  [TM_Decision_Tree]
PREDICTION JOIN
  OPENQUERY([Adventure Works DW2008R2],
    'SELECT
      [ProspectiveBuyerKey],
      [FirstName],
      [LastName],
      [EmailAddress],
      [AddressLine1],
      [AddressLine2],
      (DATEDIFF(YYYY,[BirthDate],getdate())) AS [calcAge],
      [MaritalStatus],
      [Gender],
      [YearlyIncome],
      [TotalChildren],
      [NumberChildrenAtHome],
```

```
[HouseOwnerFlag],
      [NumberCarsOwned],
      [Unknown]
    FROM
      [dbo].[ProspectiveBuyer]
    ') AS t
ON
  [TM Decision Tree].[Marital Status] =
t.[MaritalStatus] AND
  [TM_Decision_Tree].[Gender] = t.[Gender] AND
  [TM_Decision_Tree].[Yearly Income] =
t.[YearlyIncome] AND
  [TM Decision Tree].[Total Children] =
t.[TotalChildren] AND
  [TM_Decision_Tree].[Number Children At Home] =
t.[NumberChildrenAtHome] AND
  [TM Decision Tree].[House Owner Flag] =
t.[HouseOwnerFlag] AND
  [TM_Decision_Tree].[Number Cars Owned] =
t.[NumberCarsOwned] AND
  [TM_Decision_Tree].[Age] = t.[calcAge] AND
  [TM_Decision_Tree].[Bike Buyer] = t.[Unknown]
WHERE
  [TM_Decision_Tree].[Bike Buyer] =1
```



Consulta en sql server

