Homework 1

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I used for solving this exercise Matlab. For the distance I used euclidian distance. My clustering algorithm calculates for every Datavector the distances to every centroid and then checks which centroid is closed. The vector is add to the closest cluster. When all datavectors are put in a cluster the new cluster centroids are calculated.

My abort criterion is the change between the centroids. If they still change I will apply the algorithms again. Because of the computer accuracy I do not use 0, but 10^{-15} . I also let the algorithm stop after 3 rounds, but it is already good enough after 4 rounds.

After the first round all data is clustered in cluster 2 and 3. I decided to let be the first centroid (-1000, -1000, -1000). Than it does not influence the result. But two other options would be

- 1. Set the centroid to 0 or 1
- 2. random restart of just this or all cluster centroids

When I would have to choose another option I would use a restart, but it was not said, that it has to be 3 clusters, so I decided it would be okay like this.

Ta	belle 1: Distances first rou	ınd
$\operatorname{distance_to_cluster_1}$	$\operatorname{distance_to_cluster_2}$	$distance_to_cluster_3$
43000,00002	3000,002694	23000,00899
$35000,\!07642$	5000,486525	15000,09671
26888,00021	13112,00189	6888,039561
44000,00413	4000,028178	$24000,\!00035$
40000,00664	$20,\!15113235$	20000,002
47500,00003	7500,000631	$27500,\!00685$
31000,02992	9000,088284	11000,02773
45000,0265	5000,212648	$25000,\!02102$
37000,00023	3000,000006	17000,00789
15000, 23523	25000, 12893	5000,409169
45000,00443	5000,027278	$25000,\!00028$
42500,00031	2500,000209	$22500,\!00527$
44000,00319	4000,02221	24000,00046
43500,00003	3500,002355	23500,00885
25000,03222	15000,04597	5000,051895
40000,00014	7,034833476	20000,01362
$42200,\!02875$	2200,477885	$22200,\!02037$
41500,00014	$1500,\!016496$	$21500,\!01267$
33000,00058	7000,000319	13000,00815
47000,00118	7000,004581	27000,0024

Tabelle 2:	centroids after	first round
$\operatorname{centroid}_{1}$	$\operatorname{centroid}_2$	$\operatorname{centroid}_3$
-10000	8752,941176	27704
-10000	7,779844647	13,629704
-10000	22,82352941	44

Tabelle	3	Distances	second	round

distance to cluster 1	distance to cluster 2	distance to cluster 3
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$-\frac{-}{20704,03917}$
28758,56529	6247,302602	$12704,\!04405$
36006,45169	14359,07461	$4592,\!203257$
21365,89781	$2752,\!952894$	$21704,\!01257$
$24510,\!81085$	1247,073587	17704,00916
18878,30691	6252,966011	$25204,\!03083$
32284,64122	10247,08985	8704,002232
20652,19999	$3753,\!0757$	$22704,\!00565$
27003,89625	4247,084395	14704,04516
47192,82789	26247,14247	$7296,\!135236$
20631,35796	3752,941811	$22704,\!00875$
$22505,\!10872$	1253,018169	$20204,\!03122$
$21367,\!51278$	2752,941595	$21704,\!01232$
21734,0764	$2253,\!017414$	$21204,\!03843$
37765,65939	16247,07449	$2704,\!012108$
24496,12265	$1247,\!240632$	$17704,\!05272$
22765,44019	$953,\!4550448$	19904,00236
$23287,\!55072$	$253,\!8360288$	19204,0486
30483,445	8247,069583	$10704,\!05649$
19220,80594	5752,946555	24704,01792

 $\begin{array}{ccccc} Tabelle \ 4: \ centroids \ second \ round \\ centroid _1 & centroid _2 & centroid _3 \\ -10000 & 8112.5 & 25528 \\ -10000 & 7,029701813 & 15,1678105 \\ -10000 & 21,4375 & 44,25 \end{array}$

			Tat	belle 5: L	Istano	es thi	ird ro	und
distance	+0	cluster	1	dictore	to	aluata	. n	die

$distance_to_cluster_1$	$\operatorname{distance_to_cluster_2}$	$distance_to_cluster_3$
22116,19085	1112,627498	18528,04545
$28758,\!56529$	6887,73384	10528,0504
36006, 45169	$14999,\!51298$	2416,39972
21365,89781	$2112,\!51454$	$19528,\!01534$
$24510,\!81085$	1887,5148	15528,01088
18878,30691	$5612,\!522945$	$23028,\!03511$
$32284,\!64122$	$10887,\!53297$	$6528,\!001674$
$20652,\!19999$	3112,67816	$20528,\!00509$
$27003,\!89625$	$4887,\!517724$	$12528,\!05539$
47192,82789	$26887,\!5854$	$9472,\!102143$
20631,35796	3112,502093	$20528,\!01042$
$22505{,}10872$	$612,\!6238331$	18028,03664
$21367,\!51278$	$2112,\!50014$	19528,0146
21734,0764	$1612,\!589208$	19028,04449
37765,65939	16887, 51723	528,0438334
24496,12265	1887,603135	$15528,\!0622$
22765,44019	314,2232448	17728,00213
$23287,\!55072$	$388,\!0020635$	$17028,\!05672$
30483,445	8887,50778	8528,074443
19220,80594	5112,504053	$22528,\!02049$

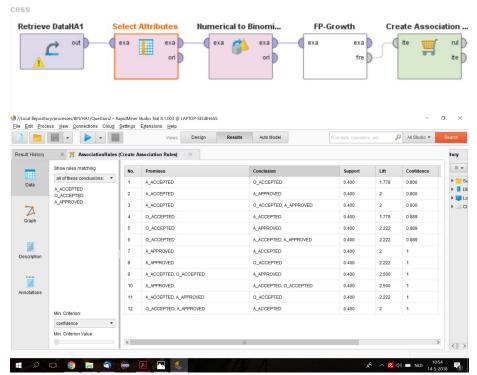
 $\begin{array}{c|cccc} Tabelle & 6: centroids third round \\ centroid _1 & centroid _2 & centroid _3 \\ -10000 & 7520 & 23822,4 \\ -10000 & 7,497431933 & 12,1369984 \\ -10000 & 22,06666667 & 37,8 \\ \end{array}$

 $\begin{array}{ccc} & \text{Tabelle 7: Cluster 1} \\ \text{amount_req} & \text{case_duration} & \text{total_activities} \end{array}$

	Tabelle 8: Cluster	
$\operatorname{amount} \operatorname{req}$	$\operatorname{case_duration}$	$total_activities$
7000	0,29309	6
15000	28,43964	74
6000	0,048206	25
10000	12,95023	26
2500	$0,\!021134$	7
5000	29,51885	46
13000	$0,\!515625$	10
5000	7,612419	25
7500	$0,\!489861$	11
6000	6,503808	22
6500	0,002049	6
10000	0,000799	3
7800	$19,\!1099$	52
8500	0,000486	3
3000	6,955382	15
10000	$0,\!000799$	3
7800	$19,\!1099$	52
8500	$0,\!000486$	3
17000	0,01375	12
3000	6,955382	15

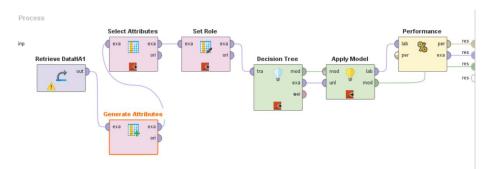
	Tabelle 9: Cluster	3
$amount_req$	$\operatorname{case_duration}$	$total_activities$
23112	$0,\!000532$	3
19000	19,78213	45
35000	19,74352	88
25000	$21{,}14506$	41
17000	0.01375	12

The process is built up like in the picture. First I selected the attributes I am assumed to consider (**Select Attributes**). Then the numerical values are translated to binomial values, because this is needed for the association rules. (**Numerical to Binomial**) RapidMiner expectes a FrequentItemSet for the **Create Association Rule**, so before I could use that I also had to use **FP-growth**.



I would pick the rule {A_ACCEPTED, O_ACCEPTED} \Rightarrow {A_APPROVED} and {A_APPROVED} \Rightarrow {A_ACCEPTED, O_ACCEPTED}, because they have the highest lift, confidence and support. When you have a closer look you will see that the sets are probably of the rules have a back- and forth relationsip, so they are symmetric. It is an interesting behaviour. Such that you could summarize it in one rule.

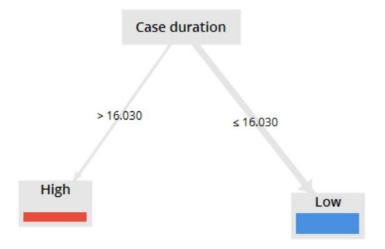
Then you could look for the next best rules, where lift, support and confidence is the highest.



The process

first changes the numerical attribute TotalActivites to a nominal one by **Generate Attributes** and **Select Attributes**. The generating consideres the old TotalActivites and contains the rule that all data, where TotalActivites is lower or the same as 40 it should be assigned to "Low" otherwise "High". The **Set Role** gives the new attribute as label, so RapidMiner knows what should have be the outcome of the Decision Tree. **Decision Tree** generates the decision Tree. Then **Apply Model** for **Performance** checking. The output is then the model and the performance of the model on the data.

The found decision tree is



If you check the Confusionmatrix

accuracy: 100.00%

	true Low	true High	class precision
pred. Low	14	0	100.00%
pred. High	0	6	100.00%
class recall	100.00%	100.00%	

you see, that this decision tree classifys the data perfectly. So you can predict by just knowing the case duration the total activities. If the case duration is higher, than also the total activities are high. This seems to be logical, if you have to do a lot this takes most of the times longer and otherwise around, if you do not need long you mostly did not do a lot of different things in the time.

1.

 $L1 = \left[\langle a,b,e,f \rangle, \langle a,b,e,c,d,b,f \rangle, \langle a,b,c,e,d,b,f \rangle, \langle a,b,c,d,e,b,f \rangle, \langle a,e,b,c,d,b,f \rangle \right]$ The α -Algorithm gives the following:

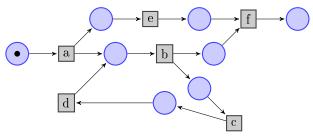
$$T_L = \{a, b, c, d, e, f\}$$

$$T_I = \{a\}$$

$$T_O = \{f\}$$

	a	b	\mathbf{c}	d	e	f
$\overline{\mathbf{a}}$	#	\rightarrow	#	#	→ # -	#
b	←	#	\rightarrow	\leftarrow		\rightarrow
\mathbf{c}	#	\leftarrow	#	\rightarrow		#
d	#	\rightarrow	\leftarrow	#		#
e	←				#	\rightarrow
f	#	\leftarrow	#	#	\leftarrow	#

$$\begin{split} X_L &= \{(\{a\},\{b\}),(\{a\},\{e\}),(\{b\},\{c\}),(\{b\},\{f\}),(\{b\},\{f,c\}),(\{c\},\{d\}),(\{d\},\{b\}),(\{e\},\{f\}),\\ Y_L &= \{(\{a\},\{b\}),(\{a\},\{e\}),(\{d\},\{b\}),(\{c\},\{d\}),(\{b\},\{f,c\})\} \end{split}$$



This model is sound, because it is bounded and live.

2.

$$L2 = \left[\langle a, b, c, d \rangle, \langle a, c, b, d \rangle, \langle a, e, f, d \rangle, \langle a, e, g, d \rangle, \langle a, e, h, d \rangle \right]$$

The α -Algorithm gives the following:

$$T_L = \{a, b, c, d, e, f, g, h\}$$

$$T_I = \{a\}$$

$$T_O = \{d\}$$

```
    a
    b
    c
    d
    e
    f
    g
    h

    a
    #
    →
    →
    #
    →
    #
    #
    #

    b
    ←
    #
    |
    →
    #
    #
    #
    #

    c
    ←
    |
    #
    →
    +
    #
    #
    #
    #

    d
    #
    ←
    ←
    #
    #
    #
    +
    ←
    ←
    ←

    e
    ←
    #
    #
    #
    #
    →
    ←
    #
    #
    #

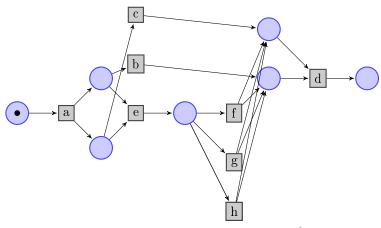
    f
    #
    #
    #
    →
    ←
    #
    #
    #
    #

    g
    #
    #
    #
    →
    ←
    #
    #
    #

    h
    #
    #
    #
    →
    ←
    #
    #
    #
```

$$\begin{split} X_L &= \{(\{a\},\{b\}),(\{a\},\{c\}),(\{a\},\{e\}),(\{a\},\{b,e\}),(\{a\},\{c,e\}),(\{b\},\{d\}),\\ &\quad (\{c\},\{d\}),(\{e\},\{f\}),(\{e\},\{g\}),(\{e\},\{h\}),(\{e\},\{f,g\}),(\{e\},\{f,h\}),(\{e\},\{g,h\}),(\{e\},\{f,g,h\}),\\ &\quad (\{g\},\{d\}),(\{h\},\{d\}),(\{f\},\{d\}),\\ &\quad (\{g,h\},\{d\}),(\{g,f\},\{d\}),(\{h,f\},\{d\}),(\{g,f,h\},\{d\}),\\ &\quad (\{g,b\},\{d\}),(\{g,c\},\{d\}),(\{h,b\},\{d\}),(\{h,c\},\{d\}),(\{f,b\},\{d\}),(\{f,c\},\{d\}),\\ &\quad (\{g,f,b\},\{d\}),(\{g,f,c\},\{d\}),(\{g,b,h\},\{d\}),(\{g,c,h\},\{d\}),(\{b,f,h\},\{d\}),(\{c,f,h\},\{d\}),\\ &\quad (\{g,f,h,b\},\{d\}),(\{g,f,h,c\},\{d\})\} \end{split}$$

$$Y_L &= \{(\{a\},\{b,e\}),(\{a\},\{c,e\}),(\{e\},\{f,g,h\}),(\{g,f,h,b\},\{d\}),(\{g,f,h,c\},\{d\})\}$$



This one is sound, because it is bounded and live. If you try every possible trace you see that easily.

3.

$$L3 = \left[\langle d, c, b, e, f \rangle, \langle a, e, f \rangle, \langle d, b, b, c, ef \rangle, \langle a, b, c, d, e, f \rangle, \langle b, d, a, c, f \rangle \right]$$

The α -Algorithm gives the following:

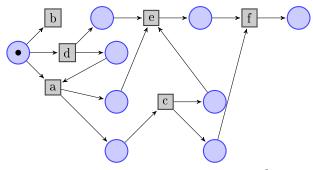
$$T_L = \{a, b, c, d, e, f\}$$

$$T_I = \{a, b, d\}$$

$$T_O = \{f\}$$

	a	b	\mathbf{c}	$^{\mathrm{d}}$	e	f
a	#	\rightarrow	\rightarrow	←	\rightarrow	#
b	←				\rightarrow	#
\mathbf{c}	←		#		\rightarrow	\rightarrow
d	#			#	\rightarrow	#
\mathbf{e}	←	\leftarrow	\leftarrow	\leftarrow	#	\rightarrow
\mathbf{f}	#	#	\leftarrow	#	\leftarrow	#

$$\begin{split} X_L &= \{(\{a\},\{c\}),(\{a\},\{e\}),(\{c\},\{e\}),(\{c\},\{f\}),(\{e\},\{f\}),(\{d\},\{a\}),(\{d\},\{e\})\} \\ Y_L &= X_L \end{split}$$

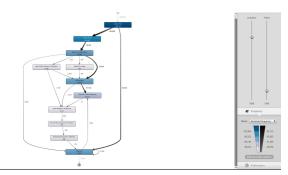


It is not live, because it is never possible to fire a, so it is neither sound.

a)

The process has 15370 cases and 561470 events. Median number of events is 5 (and average number The average duration is 48.8 weeks.

b)



The loop tells us, that a part of all people had to pay at least two times. If you check the log data, you can see, that mostly it happens in the next 30 days. It happen 4306 and for 4014 cases. So there are cases, where it happens more than one time. You also can see, that it happens at most 14 times.

c)



You can see that the distribution is going up and down a little bit, but there are 10 laces to see where more happens. Further in the end activity gets lower and the distance between the 6th and 7th lace is higher than between the others. It is always around the typical paydays, so probably a lot of people then have the money to pay the fine.

d)

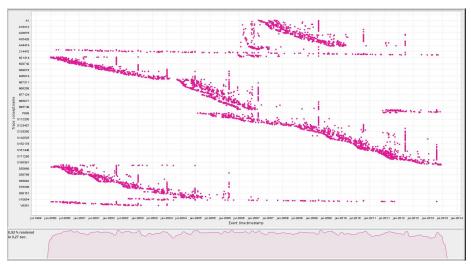
There are 231 variants. The third most frequent variant has 20385 instances. It just contains the behaviour create and send fine. Nothing more happens there.

$\mathbf{e})$

It is just possible to 43% or 56%. The average case duration shorts to 45.2 weeks from 48.8 weeks and the median to 20.9 weeks from 28.3 weeks in both cases. So the most common cases are in average faster finished than all cases in average.

The median is the instance in the middel. So if you write down all instances sorted, it is the middel one. The average is the sum of all instances divided by the number of instances. The average can change a lot for big or extree small oultiers. The median shows more a real duration in the middle of all instance durations.

f)



In the first dotted chart you see when which case is active.



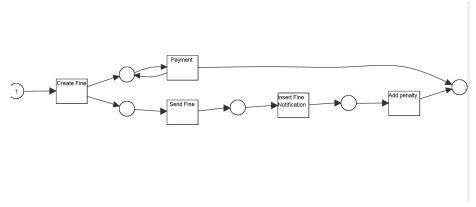
For interpreting the dot chart for the question c) I changed the y-axis to the event names so I can see when which events happening. I also zoomed in so I can see the months of the years. Now you can check better the dates of the peaks.

You can see that **Send for Credit Collection** is strongly connected to the peakes to see in c).

Payment happens mostly always, but still it is a little bit bundled at the peaks.

Furthermore I would say, that in disco it is more easy to see when peaks happen, but in prom better to see what is happen on the peak days.

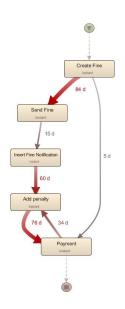
 $\mathbf{g})$



If you apply the alpha-algoirthm on the not filtered data you can not see so clear, that in the resulting chart the payment can happen always and also infinite often. It sounds weird, because you do not expect someone paying before he gets the fine, but looking at the data it happens.

Also in the filtered version the people or pay after the fine is created or pay to late, so that they get a penalty.

h)





84days is the median for send fine. I chose the median, because then you get a better idea what happens most of the times. The biggest mean is 23.3 weeks between add penalty and payment.

What you also can see, is that after $60 \mathrm{days}$ always a penalty is added.