

# Homework 1

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## Question 1

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.

Tabelle 1: Distances first round

| distance_to_cluster_1 | 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

| centroid_1 | centroid_2  | 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 |
|-----------------------|-----------------------|-----------------------|
| 22116,19085           | 1753,037892           | 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           |

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      |

Tabelle 5: Distances third round

| distance_to_cluster_1 | 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           |

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       |

Tabelle 7: Cluster 1

| amount_req | case_duration | total_activities |
|------------|---------------|------------------|
|------------|---------------|------------------|

Tabelle 8: Cluster 2

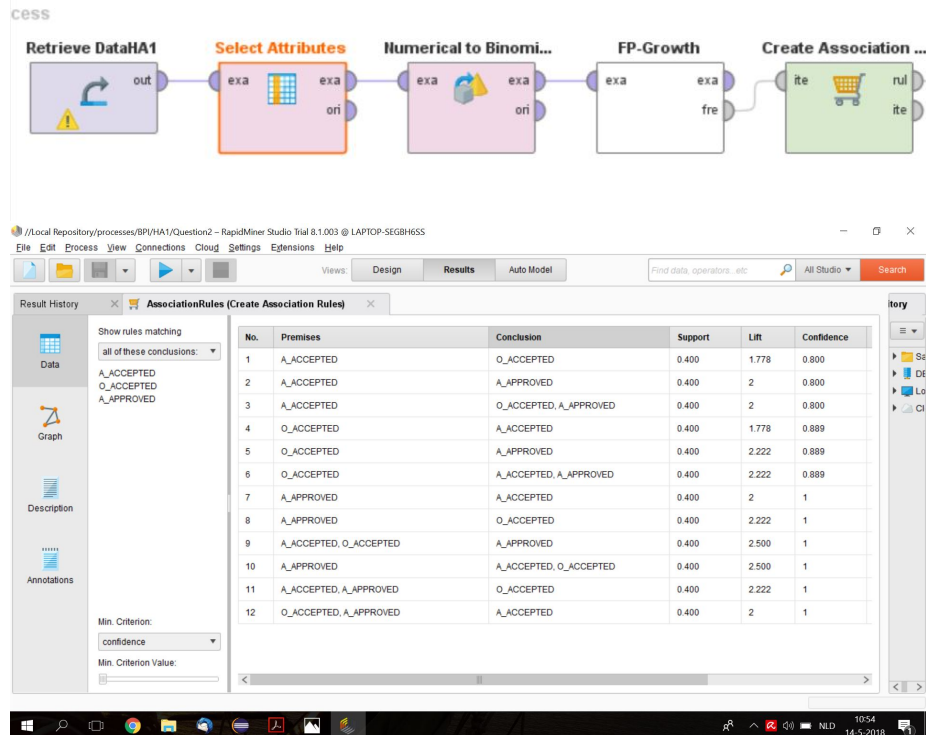
| amount_req | case_duration | total_activities |
|------------|---------------|------------------|
| 7000       | 0,29309       | 6                |
| 15000      | 28,43964      | 74               |
| 6000       | 0,048206      | 25               |
| 10000      | 12,95023      | 26               |
| 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 | case_duration | total_activities |
|------------|---------------|------------------|
| 23112      | 0,000532      | 3                |
| 19000      | 19,78213      | 45               |
| 35000      | 19,74352      | 88               |
| 25000      | 21,14506      | 41               |
| 17000      | 0,01375       | 12               |

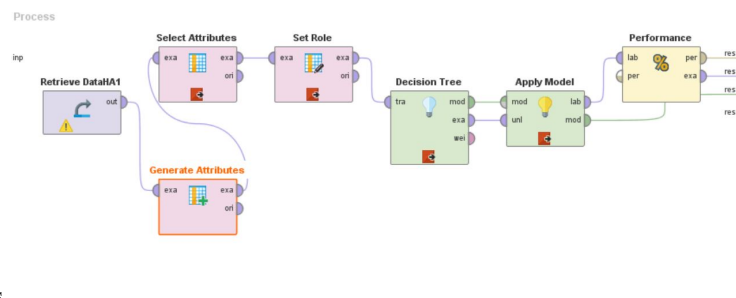
## Question2

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 expects 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 relationship. Such that you can summarize it in one rule. Then you could look for the next best rules, where lift, support and confidence is the highest.

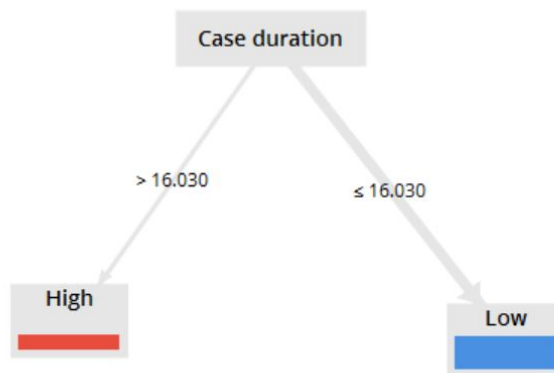
### Question 3



The process

first changes the numerical attribute TotalActivites to a nominal one by **Generate Attributes** and **Select Attributes**. The generating considers 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.



## Question 4

1.

$$L1 = [\langle a, b, e, f \rangle, \langle a, b, e, c, d, b, f \rangle, \langle a, b, c, e, d, b, f \rangle, \langle a, b, c, d, e, b, f \rangle, \langle a, e, b, c, d, b, f \rangle]$$

The  $\alpha$ -Algorithm gives the following:

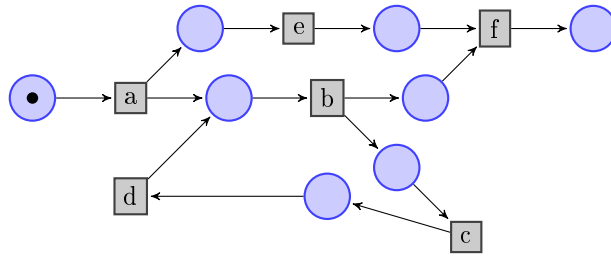
$$T_L = \{a, b, c, d, e, f\}$$

$$T_I = \{a\}$$

$$T_O = \{f\}$$

|   | a | b | c | d | e | f |
|---|---|---|---|---|---|---|
| a | # | → | # | # | → | # |
| b | ← | # | → | ← |   | → |
| c | # | ← | # | → |   | # |
| d | # | → | ← | # |   | # |
| e | ← |   |   |   | # | → |
| f | # | ← | # | # | ← | # |

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



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

2.

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

The  $\alpha$ -Algorithm gives the following:

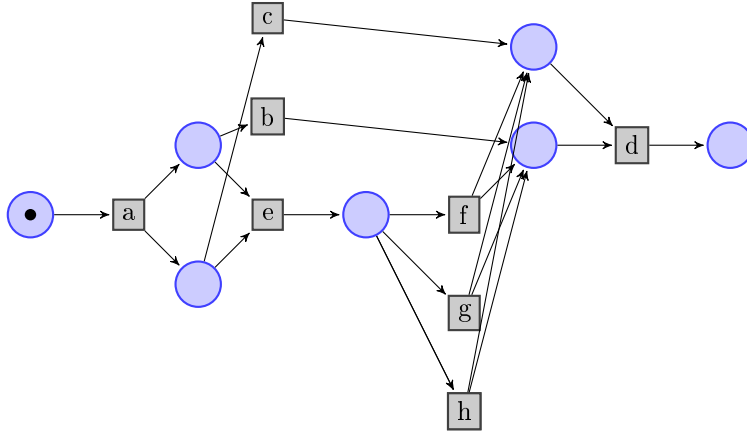
$$T_L = \{a, b, c, d, e, f, g, h\}$$

$$T_I = \{a\}$$

$$T_O = \{d\}$$

|   | a | b | c | d | e | f | g | h |
|---|---|---|---|---|---|---|---|---|
| a | # | → | → | # | → | # | # | # |
| b | ← | # |   | → | # | # | # | # |
| c | ← |   | # | → | # | # | # | # |
| d | # | ← | ← | # | # | ← | ← | ← |
| e | ← | # | # | # | # | → | → | → |
| f | # | # | # | → | ← | # | # | # |
| g | # | # | # | → | ← | # | # | # |
| h | # | # | # | → | ← | # | # | # |

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



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

3.

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

The  $\alpha$ -Algorithm gives the following:

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

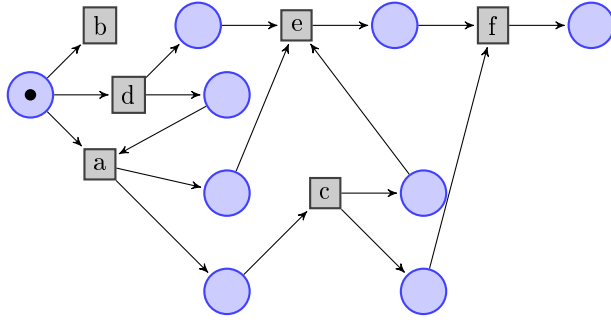
$$T_I = \{a, b, d\}$$

$$T_O = \{f\}$$

|   | a | b | c | d | e | f |
|---|---|---|---|---|---|---|
| a | # | → | → | ← | → | # |
| b | ← |   |   |   | → | # |
| c | ← |   | # |   | → | → |
| d | → |   |   | # | → | # |
| e | ← | ← | ← | ← | # | → |
| f | # | # | ← | # | ← | # |
|   | # |   |   |   |   |   |

$$X_L = \{(\{a\}, \{c\}), (\{a\}, \{e\}), (\{c\}, \{e\}), (\{c\}, \{f\}), (\{e\}, \{f\}), (\{d\}, \{a\}), (\{d\}, \{e\})\}$$

$$Y_L = X_L$$



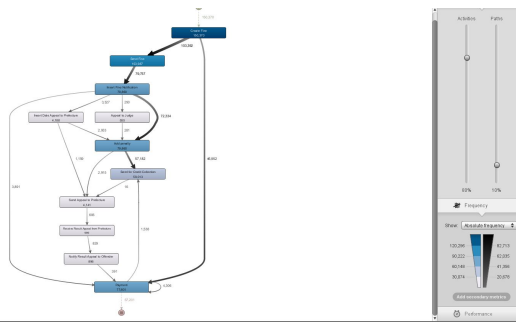
It is not live, because it is never possible to fire a, so it is neither sound.

## Question 5

a)

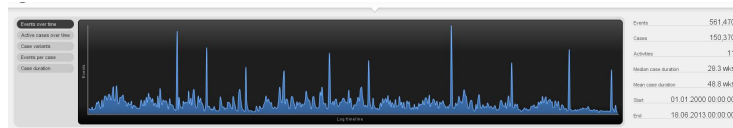
The process has 15370 cases and 561470 events. Median number of events is 5 (and average number The average duration is 48.8 weeks.

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.

e)

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

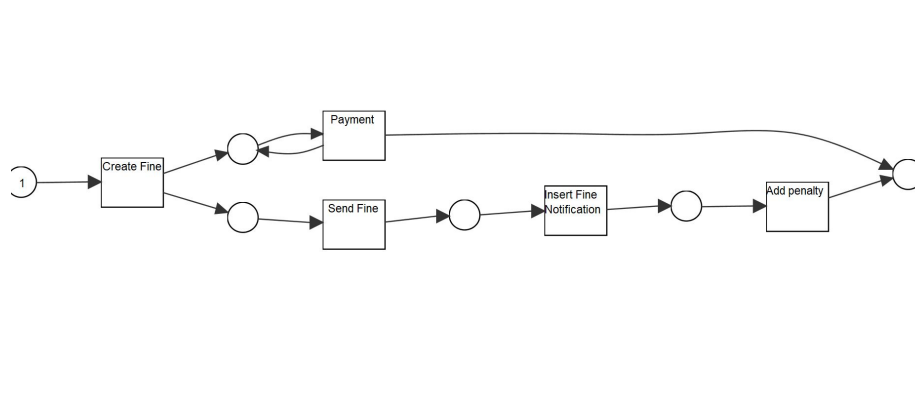
**f)**



The screenshot displays the 'Data Explorer' interface. On the left, a sidebar contains a 'Data Explorer' section with a tree view showing the data structure. The main area shows a time series plot for 'COVID-19 cases' from January 2020 to March 2021. The plot features a green line representing the data and a red line representing the trend. The y-axis is labeled 'COVID-19 cases' and ranges from 0 to 100,000. The x-axis is labeled 'Date' and shows dates from 2020-01-01 to 2021-03-01. The legend indicates 'COVID-19 cases' and 'Trend'.

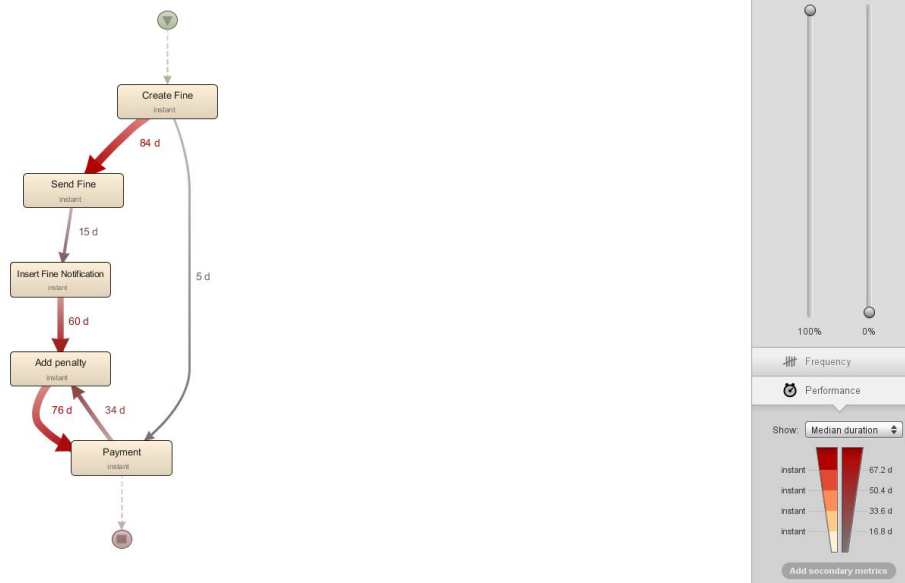
Payment happens close to always but still is a little bit bundled at the peaks. Furthermore I would say, that in disco it is more easy to see when peaks happen, but in prom better to see what is happen on the peak days.

g)



If you apply the alpha-algorithm 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 mean is 85.1 days. What you also can see, is that after 60days always a penalty is added.