

# MODULE 11

## INDUCTION AND ASSOCIATION RULE



Sort by :

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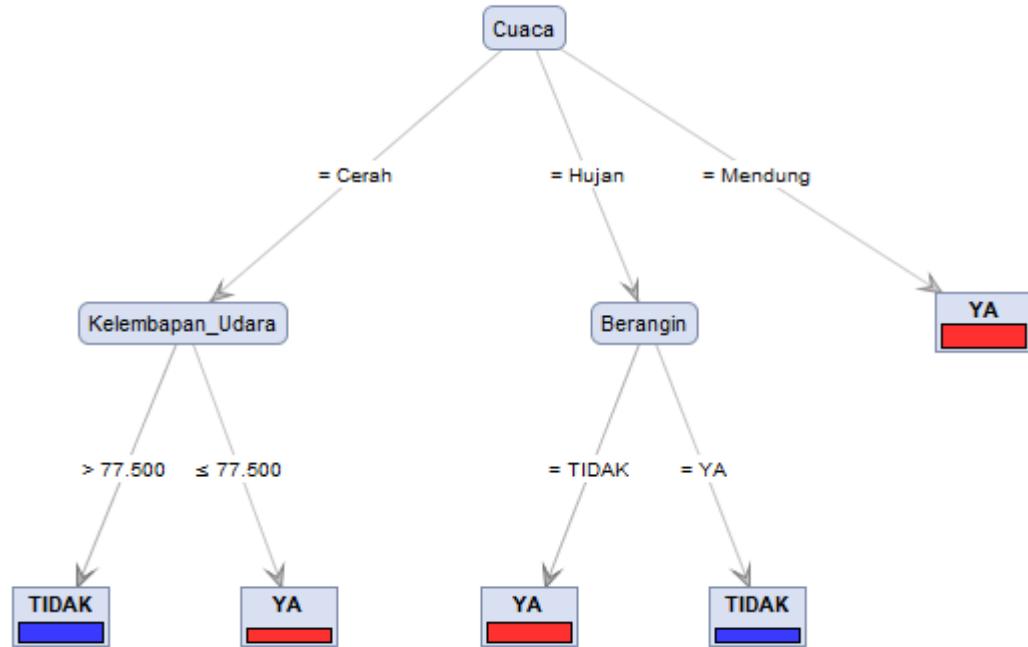
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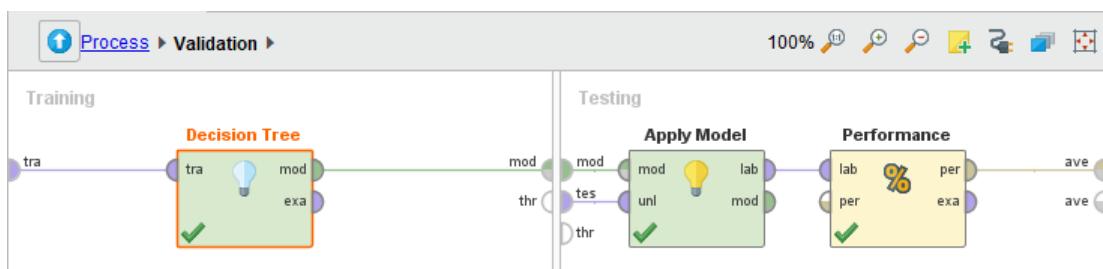
## Praktikum Steps

### Induction Rules of Data Cuaca

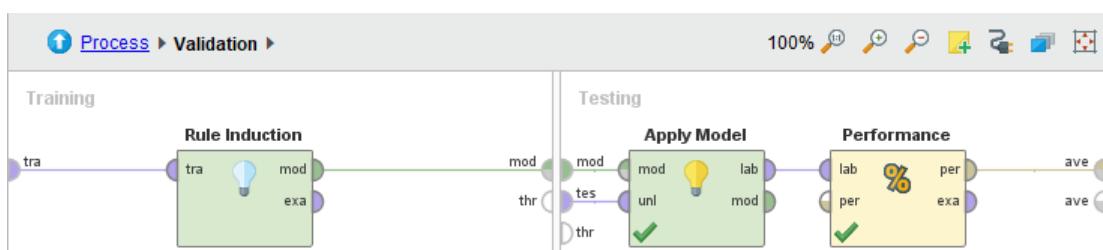
1. Use process model praktikum data mining from module 9 activity 9.4.2 (if not save, redo the activity again).
2. In module 9 activity will produce the following decision tree.



3. Open area Process – Validation.



4. Replace Decision Tree operator to Rule Induction then run the process without changing any parameters.



5. An Induction Rule will be obtained from the training data provided which is called a **RuleModel (Rule Model)**.

## RuleModel

```

if Kelembapan_Udara ≤ 82.500 then YA  (1 / 6)
if Cuaca = Cerah then TIDAK  (3 / 0)
if Cuaca = Mendung then YA  (0 / 2)
if Suhu ≤ 70.500 then YA  (0 / 1)
else TIDAK  (0 / 0)

correct: 12 out of 13 training examples.

```

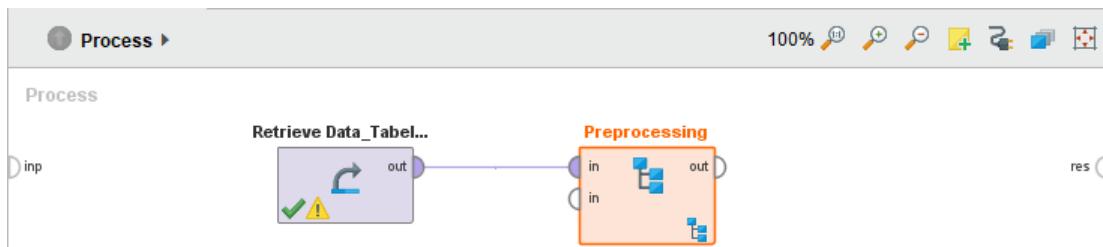
6. From the result of RuleModel we can see causality (using if... then...). The rule can be used as decision making whether someone will be playing tennis or not depends on the weather which occur.
7. Correct parameter showing the number of corrected data in the induction rules on the total data used as training data.
8. Performance of Induction Rule Model

accuracy: 65.00% +/- 47.43% (micro average: 71.43%)

	true TIDAK	true YA	class precision
pred. TIDAK	2	1	66.67%
pred. YA	3	8	72.73%
class recall	40.00%	88.89%	

### Data Cuaca Association Rule

1. Create New Process in RapidMiner.
2. Drag DataCuaca\_Training into process area.
3. Add Subprocess operator into area then rename it into **Preprocessing**. Connect retrieve output port into preprocessing input port.

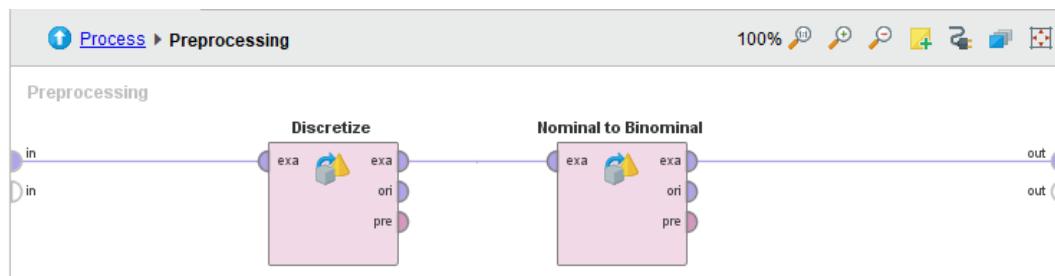


4. Double click on **Preprocessing** operator to open *Nested Chain* area then add the following operators:
- Discretize by Frequency**



Rename this operator into **DiscretizebyFrequency** and left the parameter **number of bins = 2** then connect the input port into panel **in**.

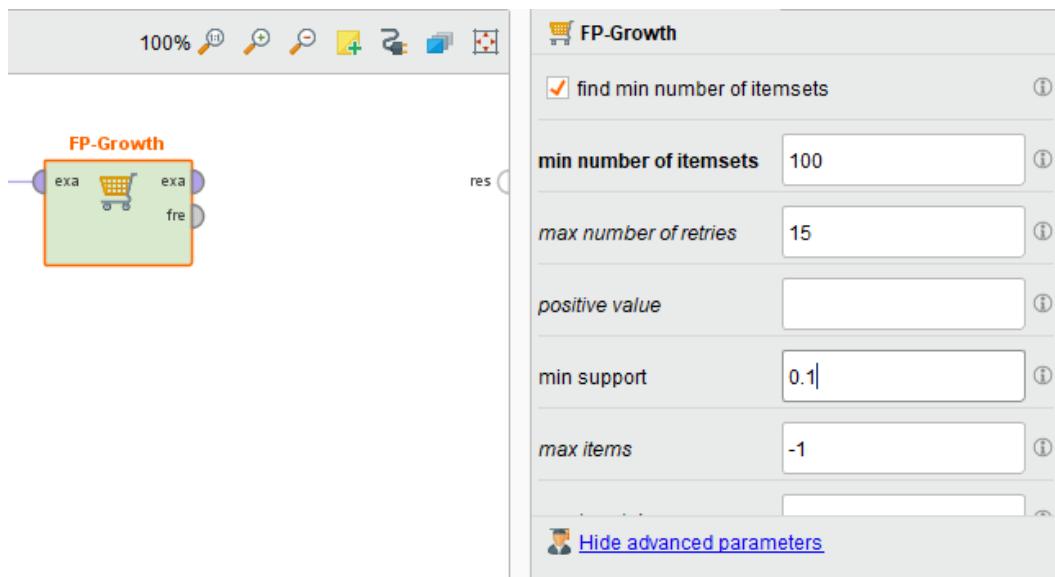
- Nominal to Binomial**



Rename this operator into **Nominal2Binomial**, connect input port with output port of **DiscretizebyFrequency** and output port with out panel.

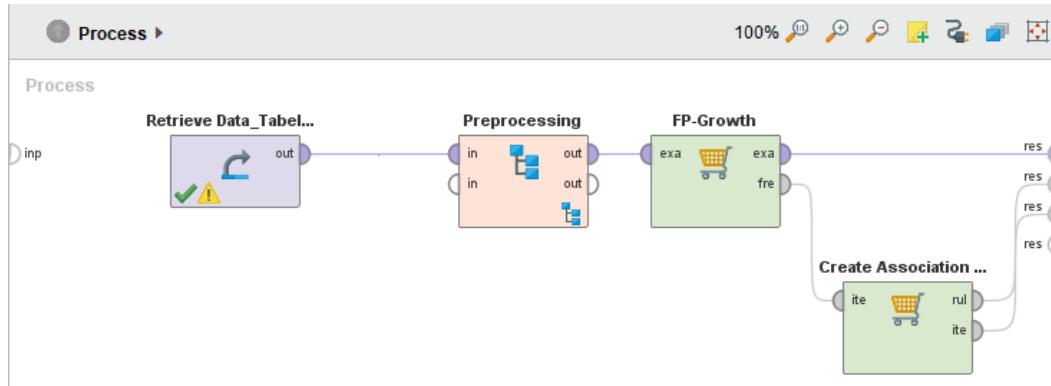
5. Back into main process, add 2 more operators:

- FP – Growth**



Change parameter **min support** into **0.1**, connect output port into with **res** connector in work area and input port into output port Preprocessing operator.

## b) Create Association Rules



Connect input port into FP – Growth fre ouput (frequent sets) and both output port into res connector in work area.

6. Run the process
7. The results of the association rules can be seen as follows:
  - a) Frequent Item Set (FP – Growth)

No. of Sets: 26	Size	Support	Item 1	Item 2	Item 3	Item 4
Total Max. Size: 4	1	0.500	Kelembapan_Udara			
Min. Size: 1	1	0.429	Suhu			
Max. Size: 4	1	0.429	Berangin			
Contains Item:	1	0.357	Cuaca = Hujan			
	1	0.357	Cuaca = Cerah			
	1	0.286	Cuaca = Mendung			
	2	0.214	Kelembapan_Udara	Suhu		
	2	0.214	Kelembapan_Udara	Berangin		
	2	0.143	Kelembapan_Udara	Cuaca = Hujan		
	2	0.214	Kelembapan_Udara	Cuaca = Cerah		
	2	0.143	Kelembapan_Udara	Cuaca = Mendung		
	2	0.143	Suhu	Berangin		
	2	0.071	Suhu	Cuaca = Hujan		
	2	0.214	Suhu	Cuaca = Cerah		
	2	0.143	Suhu	Cuaca = Mendung		

It can be seen that the number of association rules formed is 26 sets, and the total number of max sizes = 4, which consists of 4 itemset.

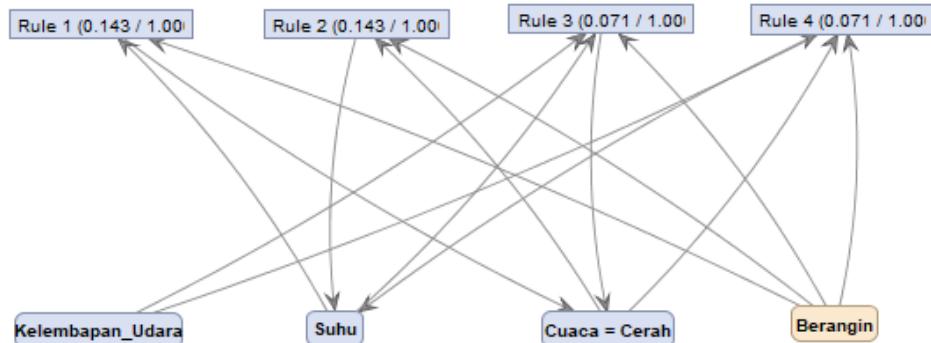
### b) Association Rules (Create Association Rules)

#### i. Table View

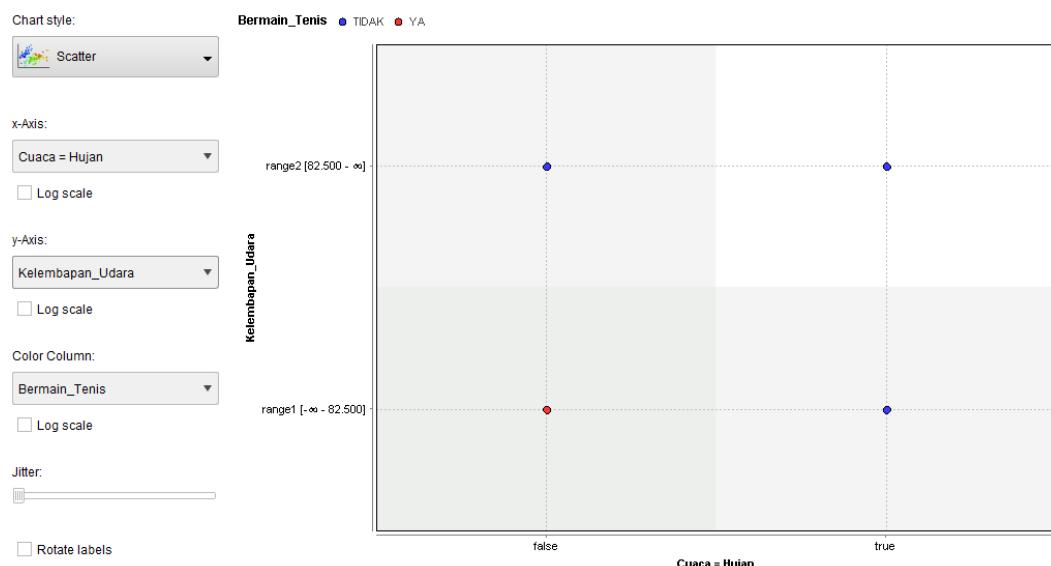
No.	Premises	Conclusion	Support	Confidence	LaPlace	Gain	p-s
1	Suhu, Berangin	Cuaca = Cerah	0.143	1	1	-0.143	0.092
2	Berangin, Cuaca = Cerah	Suhu	0.143	1	1	-0.143	0.082
3	Kelembapan_Udara, Suhu, Berangin	Cuaca = Cerah	0.071	1	1	-0.071	0.046
4	Kelembapan_Udara, Berangin, Cuaca = Cerah	Suhu	0.071	1	1	-0.071	0.041

In this table, we can see that there are 4 data pairs of premises and conclusions that show the relationship of implications based on association rules.

## ii. Graph View



## c) ExampleSet (Nominal2Binomial) Charts View



In this result it can be shown the distribution pattern of each weather data which is processed using association rules. We can change the variable in x-Axis and y-Axis and Jitter value. However, make sure that the Color Column is variable dependent, that is BERMAIN\_TENIS.

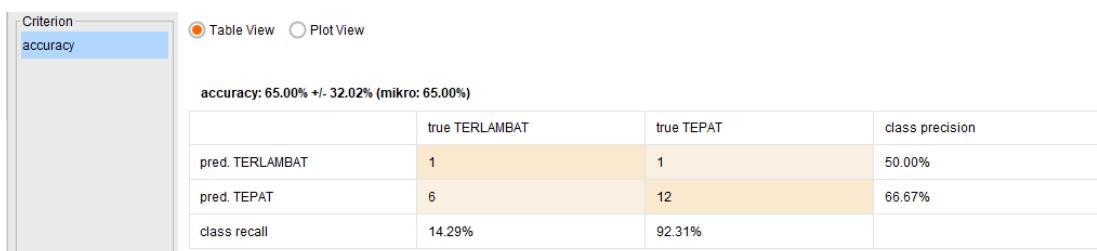
## Task (Using RapidMiner 7.1.001)

1. Induction Rule (Rule Model) and Performance Vector Value using Lama Studi Data in Task Module 6 Question No. 1

### RuleModel

```
if Asisten = YA then TEPAT  (0 / 5)
if Rerata_SKS > 19.500 then TEPAT  (0 / 2)
if Gender = PRIA and Asal_Sekolah = LUAR then TERLAMBAT  (2 / 0)
else TEPAT  (4 / 5)
```

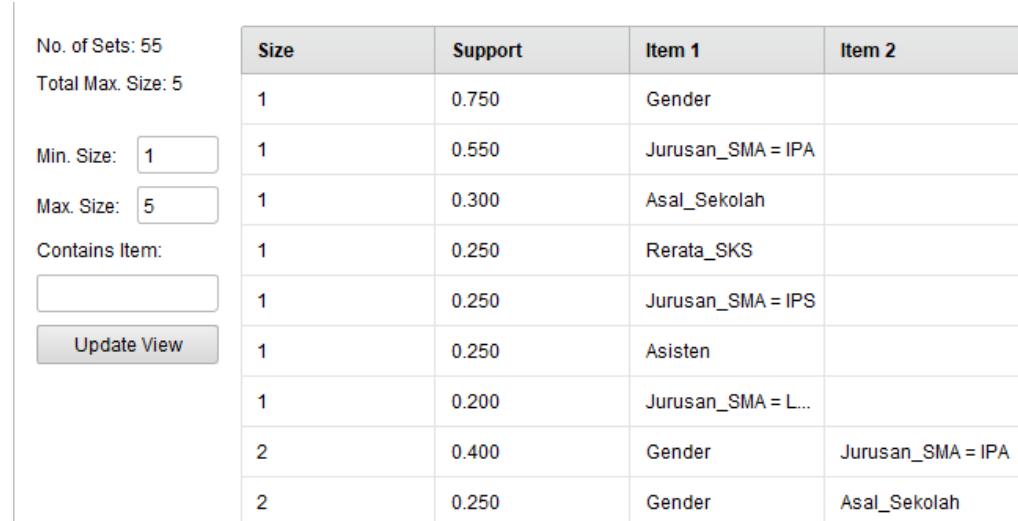
correct: 14 out of 18 training examples.



The screenshot shows the 'accuracy' performance vector in 'Table View'. The table has four columns: 'true TERLAMBAT', 'true TEPAT', 'class precision', and 'class recall'. The data rows are: pred. TERLAMBAT (1, 1, 50.00%), pred. TEPAT (6, 12, 66.67%), and class recall (14.29%, 92.31%).

	true TERLAMBAT	true TEPAT	class precision
pred. TERLAMBAT	1	1	50.00%
pred. TEPAT	6	12	66.67%
class recall	14.29%	92.31%	

2. With the same data training in no 1, with the provisions of the value operator Discretize by Frequency:
  - a. Number of Bins = 2
  - a) FP Growth (Table View)



The screenshot shows the FP Growth interface with various parameters set: No. of Sets: 55, Total Max. Size: 5, Min. Size: 1, Max. Size: 5, and Contains Item: (empty). The 'Update View' button is highlighted. The main table displays items and their supports across two bins:

No. of Sets: 55	Size	Support	Item 1	Item 2
Total Max. Size: 5	1	0.750	Gender	
Min. Size: 1	1	0.550	Jurusan_SMA = IPA	
Max. Size: 5	1	0.300	Asal_Sekolah	
Contains Item:	1	0.250	Rerata_SKS	
	1	0.250	Jurusan_SMA = IPS	
	1	0.250	Asisten	
	1	0.200	Jurusan_SMA = L...	
	2	0.400	Gender	Jurusan_SMA = IPA
	2	0.250	Gender	Asal_Sekolah

## b) Association Rules

Show rules matching  
all of these conclusions:

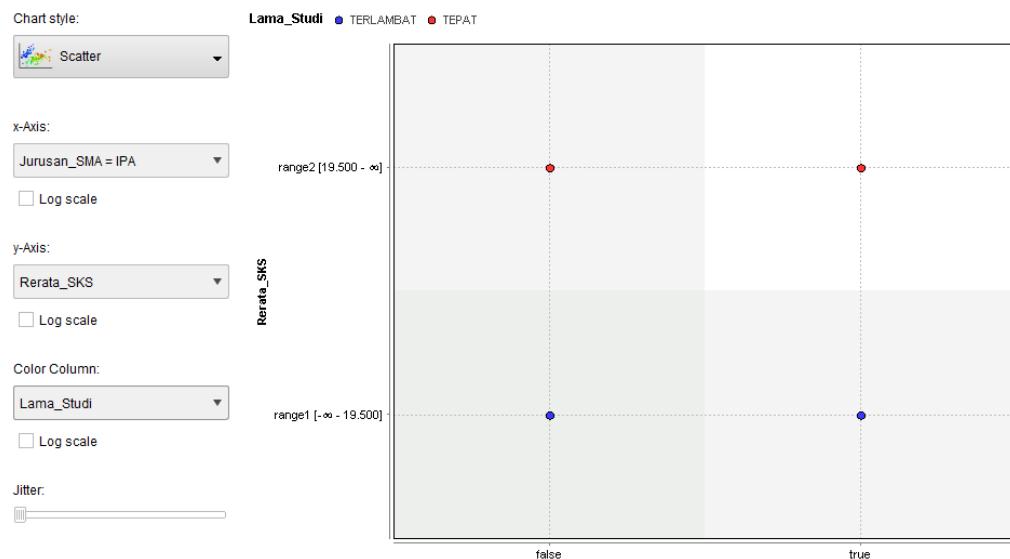
Gender  
Jurusan\_SMA = IPA  
Asal\_Sekolah  
Rerata\_SKS  
Asisten

Min. Criterion:  
confidence

Min. Criterion Value:

No.	Premises	Conclusion	Support	Confidence
1	Jurusan_SMA = IPS	Gender	0.200	0.800
2	Asisten	Gender	0.200	0.800
6	Jurusan_SMA = IPA, Rerata_SKS	Gender	0.100	1
7	Asal_Sekolah, Rerata_SKS	Gender	0.150	1
8	Asal_Sekolah, Jurusan_SMA = IPS	Gender	0.050	1
9	Asal_Sekolah, Jurusan_SMA = LAIN	Gender	0.050	1
10	Rerata_SKS, Jurusan_SMA = IPS	Gender	0.100	1
11	Rerata_SKS, Asisten	Gender	0.150	1
12	Rerata_SKS, Jurusan_SMA = LAIN	Gender	0.050	1
13	Asisten, Jurusan_SMA = LAIN	Gender	0.050	1
14	Jurusan_SMA = IPA, Rerata_SKS	Asisten	0.100	1
15	Asal_Sekolah, Jurusan_SMA = IPS	Rerata_SKS	0.050	1
16	Asal_Sekolah, Jurusan_SMA = LAIN	Rerata_SKS	0.050	1
17	Rerata_SKS, Jurusan_SMA = LAIN	Asal_Sekolah	0.050	1
18	Asal_Sekolah, Jurusan_SMA = LAIN	Asisten	0.050	1

## c) Distribution Graphic in ExampleSet



b. Number of Bins = 3

a) FP Growth (Table View)

No. of Sets: 82	Size	Support	Item 1	Item 2	Item 3	Item 4	Item 5
Total Max. Size: 5							
Min. Size:	1	0.750	Gender				
Max. Size:	5	0.550	Jurusan_SMA = ...				
Contains Item:		0.400	Rerata_SKS = r...				
		0.350	Rerata_SKS = r...				
		0.300	Asal_Sekolah				
		0.250	Rerata_SKS = r...				
		0.250	Jurusan_SMA = ...				
		0.250	Asisten				
		0.200	Jurusan_SMA = ...				
	2	0.400	Gender	Jurusan_SMA = ...			
	2	0.200	Gender	Rerata_SKS = r...			
	2	0.300	Gender	Rerata_SKS = r...			
	2	0.250	Gender	Asal_Sekolah			
	2	0.250	Gender	Rerata_SKS = r...			

b) Association Rules

Show rules matching	No.	Premises	Conclusion	Support
all of these conclusions:	4	Asal_Sekolah	Gender	0.250
Gender Jurusan_SMA = IPA Asal_Sekolah Rerata_SKS = range3 [19.500 - ∞] Asisten	5	Rerata_SKS = range2 [18.500 - 19.500]	Gender	0.300
	6	Rerata_SKS = range3 [19.500 - ∞]	Gender	0.250
	7	Jurusan_SMA = IPA, Rerata_SKS = range3 [19.500 - ∞]	Gender	0.100
	8	Rerata_SKS = range2 [18.500 - 19.500], Jurusan_SMA = IPS	Gender	0.100
	9	Rerata_SKS = range2 [18.500 - 19.500], Asisten	Gender	0.050
	10	Rerata_SKS = range2 [18.500 - 19.500], Jurusan_SMA = LAIN	Gender	0.050
	11	Asal_Sekolah, Rerata_SKS = range3 [19.500 - ∞]	Gender	0.150
	12	Asal_Sekolah, Jurusan_SMA = IPS	Gender	0.050
	13	Asal_Sekolah, Jurusan_SMA = LAIN	Gender	0.050
	14	Rerata_SKS = range3 [19.500 - ∞], Jurusan_SMA = IPS	Gender	0.100
	15	Rerata_SKS = range3 [19.500 - ∞], Asisten	Gender	0.150
	16	Rerata_SKS = range3 [19.500 - ∞], Jurusan_SMA = LAIN	Gender	0.050
	17	Asisten, Jurusan_SMA = LAIN	Gender	0.050
	18	Rerata_SKS = range1 [-∞ - 18.500], Asal_Sekolah	Jurusan_SMA = IPA	0.150

c) Distribution Graphic in ExampleSet

