

Data Mining

Home work 06

Association rules, 2x2 tables, interestingness

Aqeel Labash
Lecturer: Jaak Vilo

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First Question

For this task I implemented it on python and here is the code for it :

```
1 # coding: utf-8
2
3 # In [96]:
4
5 import itertools
6 import sys
7
8
9 # In [97]:
10
11 main = []
12 main.append(sorted(['B', 'C', 'A', 'F', 'H']))
13 main.append(sorted(['F', 'E', 'C', 'H']))
14 main.append(sorted(['E', 'D', 'B']))
15 main.append(sorted(['A', 'C', 'H', 'F']))
16 main.append(sorted(['E', 'F', 'A']))
17 main.append(sorted(['D', 'H', 'B']))
18 main.append(sorted(['E', 'C', 'F', 'B', 'D']))
19
20 main.append(sorted(['A', 'H', 'C', 'E']))
21 main.append(sorted(['G', 'A', 'E']))
22 main.append(sorted(['B', 'H', 'E']))
23
24 # In [98]:
25
26 class Core:
27     def __init__(self, name=None, occurrence=0):
28         self.name = name
29         self.occurrence = occurrence
30         self.sons = {}
31     def AddItem(self, items):
32         if len(items) <= 0:
33             return
34         if items[0] in self.sons.keys():
35             self.sons[items[0]].occurrence += 1
36             self.sons[items[0]].AddItem(items[1:])
37         else:
38             self.sons[items[0]] = Core(name=items[0],
39                                         occurrence=1)
40             self.sons[items[0]].AddItem(items[1:])
41     def printeverything(self, level=0):
42         # print
43         thespace = '——' * level
44         print(thespace + str(self.name) + ': ' + str(self.occurrence))
45         for itm in self.sons.keys():
46             self.sons[itm].printeverything(level=level+1)
47     # sys.stdout.write('\n')
```

```
48 #         print '\t' * level + repr(self.value)
49 #         for child in self.children:
50 #             child.other_name(level+1)
51
52
53 # In [99]:
54
55 nullobject = Core()
56 for item in main:
57     nullobject.AddItem(item)
58
59
60 # In [100]:
61
62 nullobject.printeverything()
63
```

And the result tree was like this : None:0

```
—A:5
——C:2
———E:1
———H:1
———F:1
———H:1
——B:1
———C:1
———F:1
———H:1
——E:2
———G:1
———F:1
—C:1
——E:1
———F:1
———H:1
—B:4
——C:1
———D:1
———E:1
———F:1
——E:1
———H:1
——D:2
———H:1
———E:1
```

The following picture is what (me & faiz) draw on the board:

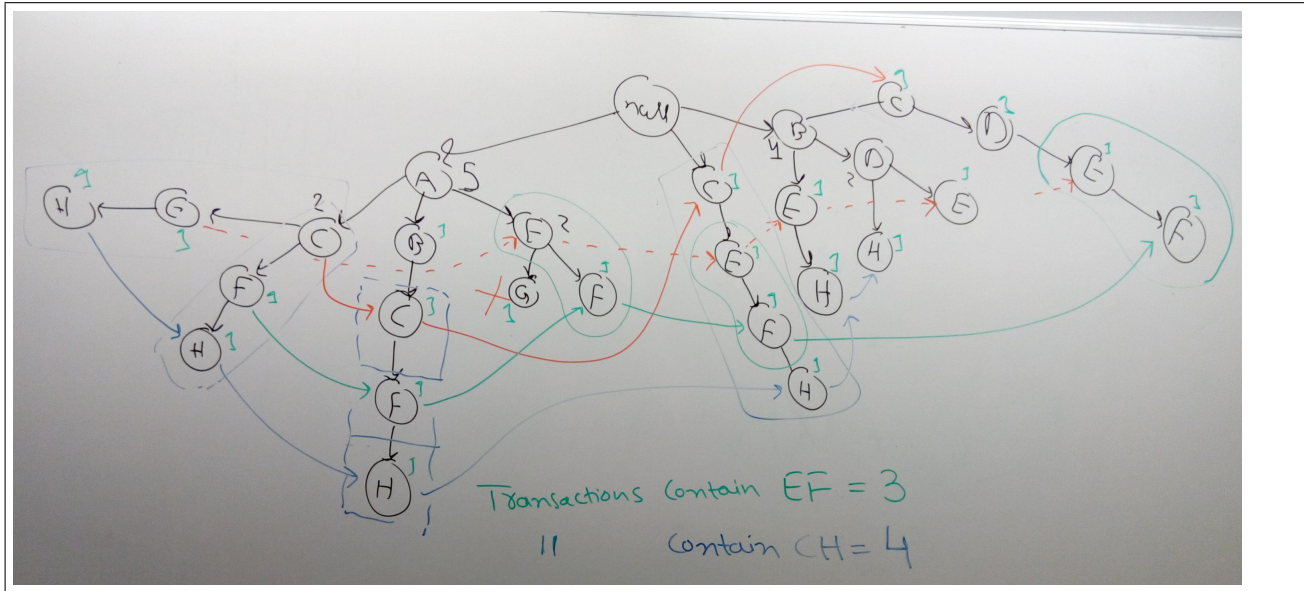


Figure 1: Image show the elements graph

How we build it:

1. Put null node at the start.
2. start adding transactions one by one
3. when the an element of the transaction already exist I just increase it's occurrence number and pass the rest of the transaction to it and so on.

The number of transaction that contain EF = 3.

The number of transaction that contain CH = 4.

Second Question

For this I question I used python and here is the code :

```

1
2 # coding: utf-8
3
4 # In[462]:
5
6 import numpy as np
7 import itertools
8 from math import sqrt
9 rnd.seed(4)
10
11
12 # In[463]:
13
14 class AllNeeded:
15     def __init__(self, value):
16         self.f11=value[0]
17         self.f10=value[1]
18         self.f01=value[2]
19         self.f00=value[3]
20         self.f1plus= self.f11+self.f10
21         self.f0plus= self.f01+self.f00
22         self.fplus1= self.f11+self.f01
23         self.fplus0=self.f10+self.f00
24         #self.oddsratio = float(self.f11*self.f00)/self.isZero( float(self.f10*self.f01))
25         self.all_confidence = min(float(self.f11)/float(self.f1plus), float(self.f11)/float(self.fplus1))
26         self.kappa = float(10000*self.f11+10000*self.f00-self.f1plus*self.fplus1-
27                             self.f0plus*self.fplus0) /float(100000000-self.f1plus*self.fplus1-self.f0plus*self.fplus0)
28         self.interest = float(10000*self.f11)/float(self.f1plus*self.fplus1)
29         self.cosine= float(self.f11)/float(sqrt(self.f1plus*self.fplus1))

```

```

30 self.Jaccard = float(self.f11)/float(self.f1plus+self.fplus1-self.f11)
31
32 def isZero(self,value):
33 return 1 if value==0 else value
34 def __str__(self):
35 return 'f11:{},f10:{},f01:{},f00:{},f1+:{},f0+:{},f+1:{},f+0:{} '.format(self.f11,self.f10,self
    .f01,
36 self.f00,self.f1plus,self.f0plus,
37 self.fplus1,self.fplus0)
38
39
40
41
42 # To generate 4 numbers sum to 10000 I used
43 # #Create the array
44 #     #result =np.random.dirichlet(np.ones(4),size=1)
45 # #multilply with 10,000 (so The sum is 10,000)
46 #     #result = result *10000
47 # #Convert the array to type int
48 #     result =np.array( result ,dtype=int)
49 # # add 2 to one of the array elements to recover the loaset precesion
50 #     result[0,np.random.randint(4)]+=2
51 #
52
53 # In[464]:
54
55 lst = []
56 total =np.array(np.random.dirichlet(np.ones(4),size=10000)*10000,dtype=int)
57 for result in total:
58 result[np.random.randint(4,size=1)]+=(10000-np.sum(result))
59 lst.append(AllNeeded(result))
60
61
62 # In[478]:
63
64 allconf = sorted(lst,key=lambda x:x.all_confidence ,reverse=True)[:10]
65 print '====All Confidence===='
66 for element in allconf:
67 print element ,',all_confidence :',element.all_confidence
68
69 print '==== kappa ====='
70 allconf = sorted(lst,key=lambda x:x.kappa ,reverse=True)[:10]
71 for element in allconf:
72 print element ,',kappa :',element.kappa
73 print '==== interest ====='
74 allconf = sorted(lst,key=lambda x:x.interest ,reverse=True)[:10]
75 for element in allconf:
76 print element ,',interest :',element.interest
77 print '==== cosine ====='
78 allconf = sorted(lst,key=lambda x:x.cosine ,reverse=True)[:10]
79 for element in allconf:
80 print element ,',cosine :',element.cosine
81 print '==== Jaccard ====='
82 allconf = sorted(lst,key=lambda x:x.Jaccard ,reverse=True)[:10]
83 for element in allconf:
84 print element ,',Jaccard :',element.Jaccard

```

and the output as following :

```

=====All Confidence===== f11:8802,f10:31,f01:21,f00:1146,f1+:8833,f0+:1167,f+1:8823,f+0:1177
,all_confidence : 0.996490433601
f11:7915,f10:11,f01:49,f00:2025,f1+:7926,f0+:2074,f+1:7964,f+0:2036 ,all_confidence : 0.993847312908
f11:8185,f10:9,f01:72,f00:1734,f1+:8194,f0+:1806,f+1:8257,f+0:1743 ,all_confidence : 0.991280125954
f11:4120,f10:45,f01:38,f00:5797,f1+:4165,f0+:5835,f+1:4158,f+0:5842 ,all_confidence : 0.989195678271
f11:7964,f10:20,f01:123,f00:1893,f1+:7984,f0+:2016,f+1:8087,f+0:1913 ,all_confidence : 0.984790404353
f11:3193,f10:52,f01:24,f00:6731,f1+:3245,f0+:6755,f+1:3217,f+0:6783 ,all_confidence : 0.983975346687
f11:6363,f10:115,f01:39,f00:3483,f1+:6478,f0+:3522,f+1:6402,f+0:3598 ,all_confidence : 0.982247607286
f11:8778,f10:169,f01:23,f00:1030,f1+:8947,f0+:1053,f+1:8801,f+0:1199 ,all_confidence : 0.981110986923
f11:8308,f10:197,f01:6,f00:1489,f1+:8505,f0+:1495,f+1:8314,f+0:1686 ,all_confidence : 0.976837154615
f11:6308,f10:162,f01:17,f00:3513,f1+:6470,f0+:3530,f+1:6325,f+0:3675 ,all_confidence : 0.974961360124
===== kappa =====
f11:4120,f10:45,f01:38,f00:5797,f1+:4165,f0+:5835,f+1:4158,f+0:5842 ,kappa : 0.982919652812
f11:3193,f10:52,f01:24,f00:6731,f1+:3245,f0+:6755,f+1:3217,f+0:6783 ,kappa : 0.982625263279
f11:7915,f10:11,f01:49,f00:2025,f1+:7926,f0+:2074,f+1:7964,f+0:2036 ,kappa : 0.981625906394

```

```

f11:8802,f10:31,f01:21,f00:1146,f1+:8833,f0+:1167,f+1:8823,f+0:1177 ,kappa : 0.974870585934
f11:8185,f10:9,f01:72,f00:1734,f1+:8194,f0+:1806,f+1:8257,f+0:1743 ,kappa : 0.972254842763
f11:5187,f10:7,f01:159,f00:4647,f1+:5194,f0+:4806,f+1:5346,f+0:4654 ,kappa : 0.966710619345
f11:6363,f10:115,f01:39,f00:3483,f1+:6478,f0+:3522,f+1:6402,f+0:3598 ,kappa : 0.966416380014
f11:3435,f10:105,f01:54,f00:6406,f1+:3540,f0+:6460,f+1:3489,f+0:6511 ,kappa : 0.965122308824
f11:2850,f10:129,f01:27,f00:6994,f1+:2979,f0+:7021,f+1:2877,f+0:7123 ,kappa : 0.962335974982
f11:2266,f10:93,f01:44,f00:7597,f1+:2359,f0+:7641,f+1:2310,f+0:7690 ,kappa : 0.961722669847
===== interest =====
f11:165,f10:312,f01:154,f00:9369,f1+:477,f0+:9523,f+1:319,f+0:9681 ,interest : 10.8436347864
f11:478,f10:185,f01:201,f00:9136,f1+:663,f0+:9337,f+1:679,f+0:9321 ,interest : 10.6180457909
f11:304,f10:118,f01:444,f00:9134,f1+:422,f0+:9578,f+1:748,f+0:9252 ,interest : 9.63073725828
f11:328,f10:610,f01:58,f00:9004,f1+:938,f0+:9062,f+1:386,f+0:9614 ,interest : 9.05907177657
f11:137,f10:939,f01:16,f00:8908,f1+:1076,f0+:8924,f+1:153,f+0:9847 ,interest : 8.32179216172
f11:415,f10:440,f01:176,f00:8969,f1+:855,f0+:9145,f+1:591,f+0:9409 ,interest : 8.21286153907
f11:932,f10:14,f01:279,f00:8775,f1+:946,f0+:9054,f+1:1211,f+0:8789 ,interest : 8.13543225158
f11:216,f10:945,f01:13,f00:8826,f1+:1161,f0+:8839,f+1:229,f+0:9771 ,interest : 8.12430181781
f11:984,f10:73,f01:293,f00:8650,f1+:1057,f0+:8943,f+1:1277,f+0:8723 ,interest : 7.29002829331
f11:716,f10:402,f01:185,f00:8697,f1+:1118,f0+:8882,f+1:901,f+0:9099 ,interest : 7.10798377474
===== cosine =====
f11:8802,f10:31,f01:21,f00:1146,f1+:8833,f0+:1167,f+1:8823,f+0:1177 ,cosine : 0.997054985476
f11:7915,f10:11,f01:49,f00:2025,f1+:7926,f0+:2074,f+1:7964,f+0:2036 ,cosine : 0.996226888987
f11:8185,f10:9,f01:72,f00:1734,f1+:8194,f0+:1806,f+1:8257,f+0:1743 ,cosine : 0.995083583875
f11:7964,f10:20,f01:123,f00:1893,f1+:7984,f0+:2016,f+1:8087,f+0:1913 ,cosine : 0.991122340845
f11:4120,f10:45,f01:38,f00:5797,f1+:4165,f0+:5835,f+1:4158,f+0:5842 ,cosine : 0.990027984416
f11:8778,f10:169,f01:23,f00:1030,f1+:8947,f0+:1053,f+1:8801,f+0:1199 ,cosine : 0.98921535114
f11:3193,f10:52,f01:24,f00:6731,f1+:3245,f0+:6755,f+1:3217,f+0:6783 ,cosine : 0.988248212585
f11:6363,f10:115,f01:39,f00:3483,f1+:6478,f0+:3522,f+1:6402,f+0:3598 ,cosine : 0.988060679228
f11:8308,f10:197,f01:6,f00:1489,f1+:8505,f0+:1495,f+1:8314,f+0:1686 ,cosine : 0.98799402648
f11:6308,f10:162,f01:17,f00:3513,f1+:6470,f0+:3530,f+1:6325,f+0:3675 ,cosine : 0.986073481348
===== Jaccard =====
f11:8802,f10:31,f01:21,f00:1146,f1+:8833,f0+:1167,f+1:8823,f+0:1177 ,Jaccard : 0.994126948272
f11:7915,f10:11,f01:49,f00:2025,f1+:7926,f0+:2074,f+1:7964,f+0:2036 ,Jaccard : 0.992476489028
f11:8185,f10:9,f01:72,f00:1734,f1+:8194,f0+:1806,f+1:8257,f+0:1743 ,Jaccard : 0.990200822647
f11:7964,f10:20,f01:123,f00:1893,f1+:7984,f0+:2016,f+1:8087,f+0:1913 ,Jaccard : 0.982360922659
f11:4120,f10:45,f01:38,f00:5797,f1+:4165,f0+:5835,f+1:4158,f+0:5842 ,Jaccard : 0.980252200809
f11:8778,f10:169,f01:23,f00:1030,f1+:8947,f0+:1053,f+1:8801,f+0:1199 ,Jaccard : 0.978595317726
f11:3193,f10:52,f01:24,f00:6731,f1+:3245,f0+:6755,f+1:3217,f+0:6783 ,Jaccard : 0.976751300092
f11:6363,f10:115,f01:39,f00:3483,f1+:6478,f0+:3522,f+1:6402,f+0:3598 ,Jaccard : 0.976369495166
f11:8308,f10:197,f01:6,f00:1489,f1+:8505,f0+:1495,f+1:8314,f+0:1686 ,Jaccard : 0.976148513688
f11:6308,f10:162,f01:17,f00:3513,f1+:6470,f0+:3530,f+1:6325,f+0:3675 ,Jaccard : 0.972406351164
More explanation about those method and how I generated them in the ipython file attached with home work
or you can find it on github,Q2.ipython

```

Third Question

For this question I used the previous code and added the following code :

```

1  x = range(1000)
2  labels = []
3  plotHandles = []
4  y1 = [AllNeeded((i,250,250,250)) for i in x]
5  #for i in range(1, num_plots + 1):
6  t, = plt.plot(x, [y.all_confidence for y in y1]) #need the ',' per ** below
7  plotHandles.append(t)
8  labels.append('All Confidence')
9  t, = plt.plot(x, [y.kappa for y in y1]) #need the ',' per ** below
10 plotHandles.append(t)
11 labels.append('kappa')
12 t, = plt.plot(x, [y.interest for y in y1]) #need the ',' per ** below
13 plotHandles.append(t)
14 labels.append('interest')
15 t, = plt.plot(x, [y.cosine+0.1 for y in y1]) #need the ',' per ** below
16 plotHandles.append(t)
17 labels.append('cosine + 0.1')

```

```

18 t, = plt.plot(x, [y.Jaccard for y in y1]) #need the ',' per ** below
19 plotHandles.append(t)
20 labels.append('Jaccard')
21 plt.legend(plotHandles, labels, 'upper left',loc=1)
22 plt.show()
23
24 labels = []
25 plotHandles = []
26 y1 = [AllNeeded((250,i,250,250)) for i in x]
27 #for i in range(1, num_plots + 1):
28 t, = plt.plot(x, [y.all_confidence for y in y1]) #need the ',' per ** below
29 plotHandles.append(t)
30 labels.append('All Confidence')
31 t, = plt.plot(x, [y.kappa for y in y1]) #need the ',' per ** below
32 plotHandles.append(t)
33 labels.append('kappa')
34 t, = plt.plot(x, [y.interest for y in y1]) #need the ',' per ** below
35 plotHandles.append(t)
36 labels.append('interest')
37 t, = plt.plot(x, [y.cosine for y in y1]) #need the ',' per ** below
38 plotHandles.append(t)
39 labels.append('cosine')
40 t, = plt.plot(x, [y.Jaccard for y in y1]) #need the ',' per ** below
41 plotHandles.append(t)
42 labels.append('Jaccard')
43 plt.legend(plotHandles, labels, 'upper left',loc=1)
44 plt.show()
45
46 labels = []
47 plotHandles = []
48 y1 = [AllNeeded((250,250,i,250)) for i in x]
49 #for i in range(1, num_plots + 1):
50 t, = plt.plot(x, [y.all_confidence for y in y1]) #need the ',' per ** below
51 plotHandles.append(t)
52 labels.append('All Confidence')
53 t, = plt.plot(x, [y.kappa for y in y1]) #need the ',' per ** below
54 plotHandles.append(t)
55 labels.append('kappa')
56 t, = plt.plot(x, [y.interest for y in y1]) #need the ',' per ** below
57 plotHandles.append(t)
58 labels.append('interest')
59 t, = plt.plot(x, [y.cosine for y in y1]) #need the ',' per ** below
60 plotHandles.append(t)
61 labels.append('cosine')
62 t, = plt.plot(x, [y.Jaccard for y in y1]) #need the ',' per ** below
63 plotHandles.append(t)
64 labels.append('Jaccard')
65 plt.legend(plotHandles, labels, 'upper left',loc=1)
66 plt.show()
67
68 labels = []
69 plotHandles = []
70 y1 = [AllNeeded((250,250,250,i)) for i in x]
71 #for i in range(1, num_plots + 1):
72 t, = plt.plot(x, [y.all_confidence for y in y1]) #need the ',' per ** below
73 plotHandles.append(t)
74 labels.append('All Confidence')
75 t, = plt.plot(x, [y.kappa for y in y1]) #need the ',' per ** below
76 plotHandles.append(t)
77 labels.append('kappa')
78 t, = plt.plot(x, [y.interest for y in y1]) #need the ',' per ** below
79 plotHandles.append(t)
80 labels.append('interest')
81 t, = plt.plot(x, [y.cosine+0.1 for y in y1]) #need the ',' per ** below
82 plotHandles.append(t)
83 labels.append('cosine+0.1')
84 t, = plt.plot(x, [y.Jaccard for y in y1]) #need the ',' per ** below
85 plotHandles.append(t)
86 labels.append('Jaccard')
87 plt.legend(plotHandles, labels, 'upper left',loc=1)
88 plt.show()
89

```

The previous code generate 4 plots. each plot represent how the change in the value in one of(f11,f10,f01,f00) would affect the measurements.

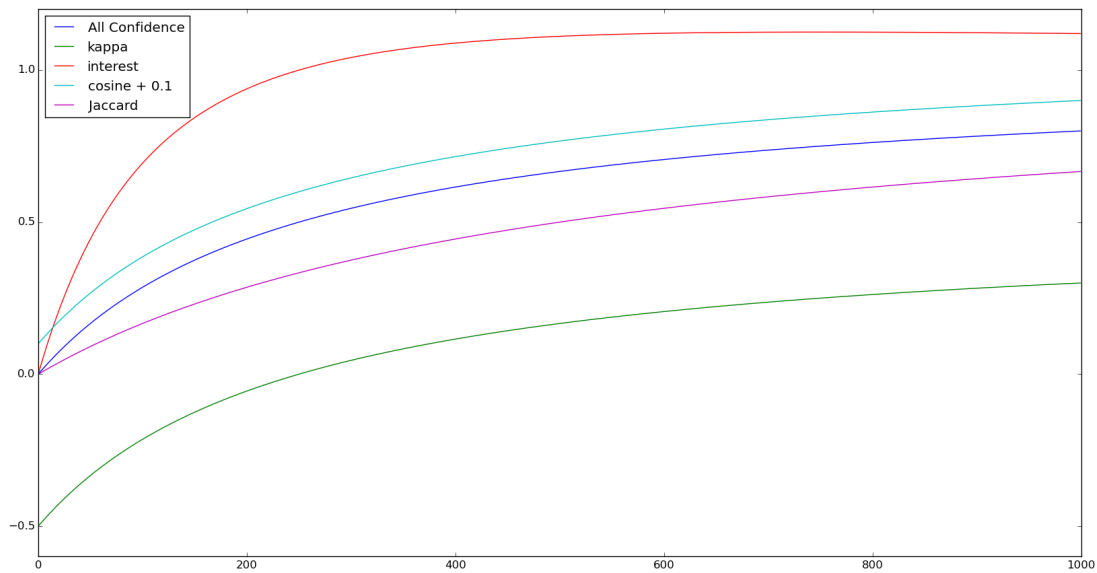


Figure 2: The measurements while changing f11

In this figure I believe all the the measurements have a logarithmic growth when we change f11.

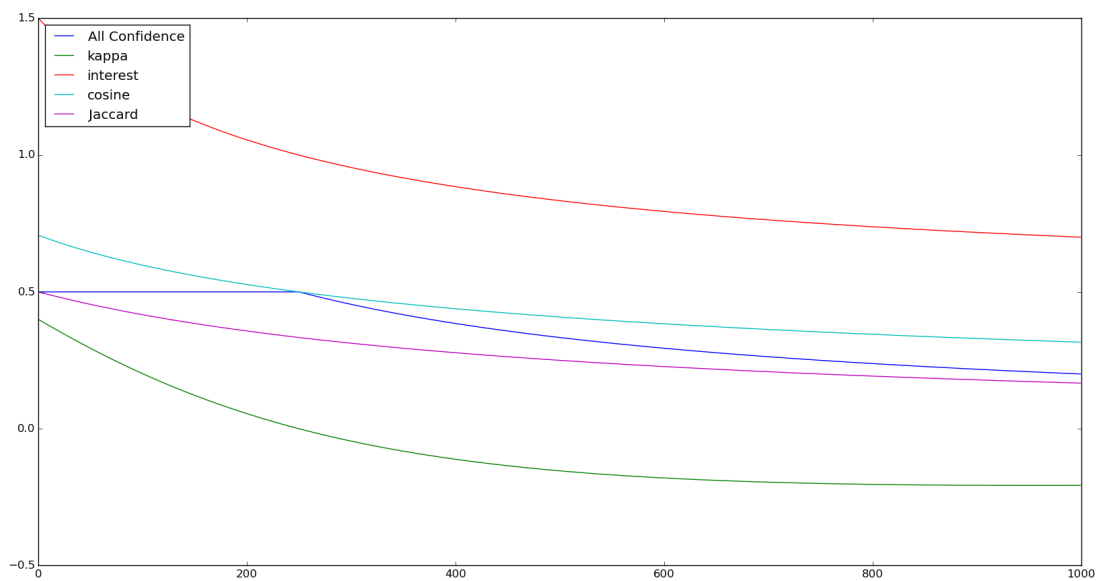


Figure 3: The measurements while changing f10

In figure 3 I think we have -log growth for all the measurements

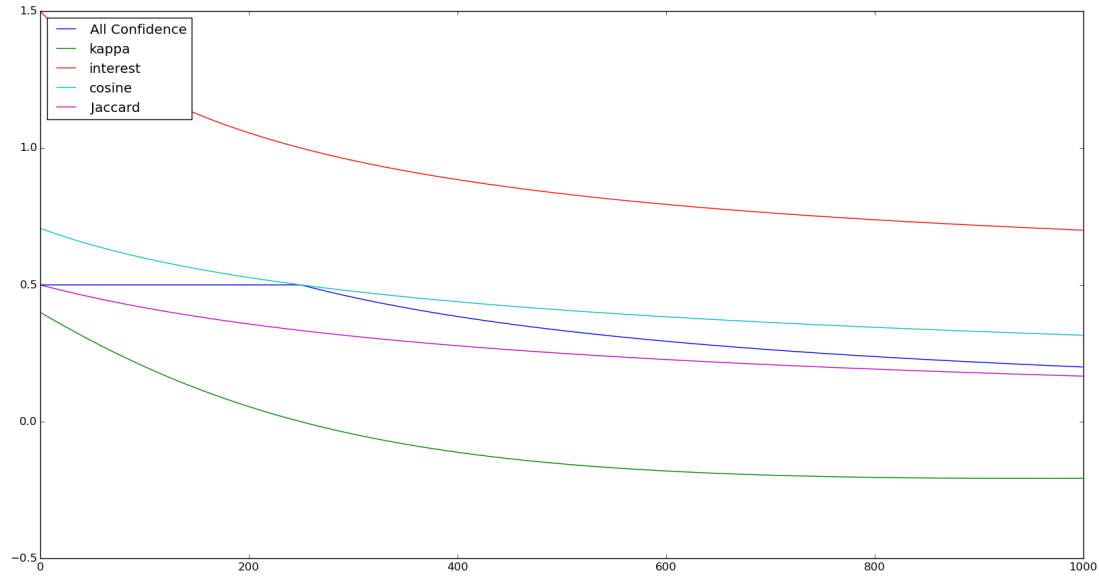


Figure 4: The measurements while changing f01

In figure 4 I would say they have same behavior as figure 3 except All confidence

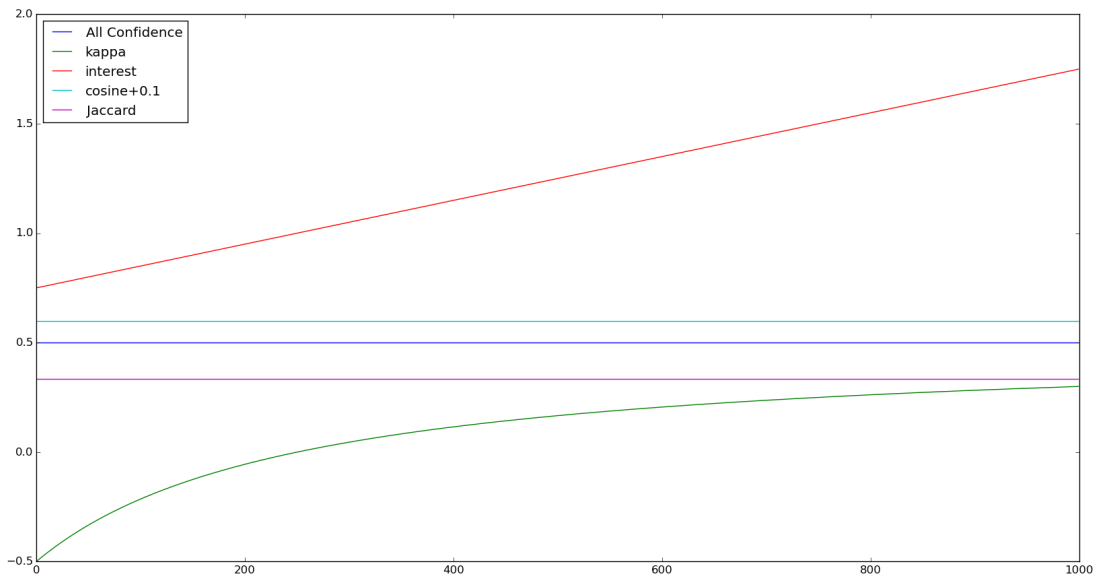


Figure 5: The measurements while changing f00

In figure 5 I think we can see that interest is linear with f00 , kappa is logarithmic, the rest doesn't get affected at all.

Fourth Question

For this question I think the code is explained so maybe I'll explain main points.

1. **apriori.** This method get all possible rules if it's used with default values. We can specify the support, confidence , lhs (left hand side : refer to X) ,rhs (right hand side: refer to Y) to get more custom rules.

2. **inspect** extract the rules list right hand and left hand.
3. **appearance or APappearance**: used to specify rhs,lhs. where each of them can be one or list of values.
4. **parameter or ASparameter**:used to specify minimum and maximum number in the item set , confidence , support , etc..

The output of the code is the following figure :

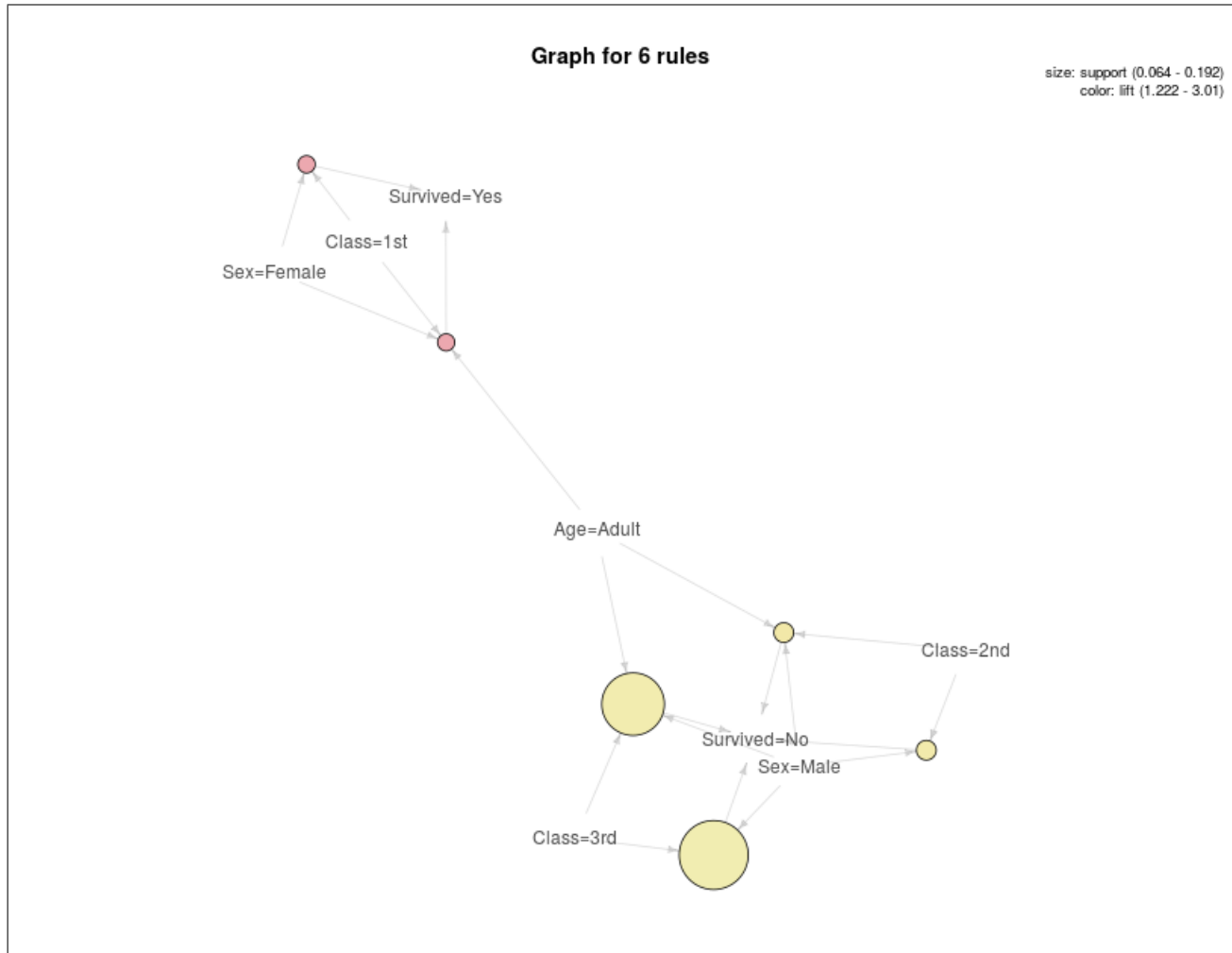


Figure 6: Shows the rules with support and confidence

In the previous figure we can notice :

1. circles : represent rule operation , has incoming arrows represent lhs , and outgoing arrows represent rhs. The color of the circle represent the lift. The size represent the support.
2. Arrows : which are directed link between different components.
3. text: which represent the rule X or Y.

So from figure 6 we can notice that

(Class=3rd,Sex=Male=> Survived =NO)

(Class=3rd,Age=Adult,Sex=Male=> Survived =NO)

both those two rules has the highest support.but not much lift. on the other side we have :

,(Class=1st,Sex=Female=> Survived =Yes)

,(Class=1st,Age=Adult,Sex=Female=> Survived =Yes)

have high lift but little support.

Fifth Question

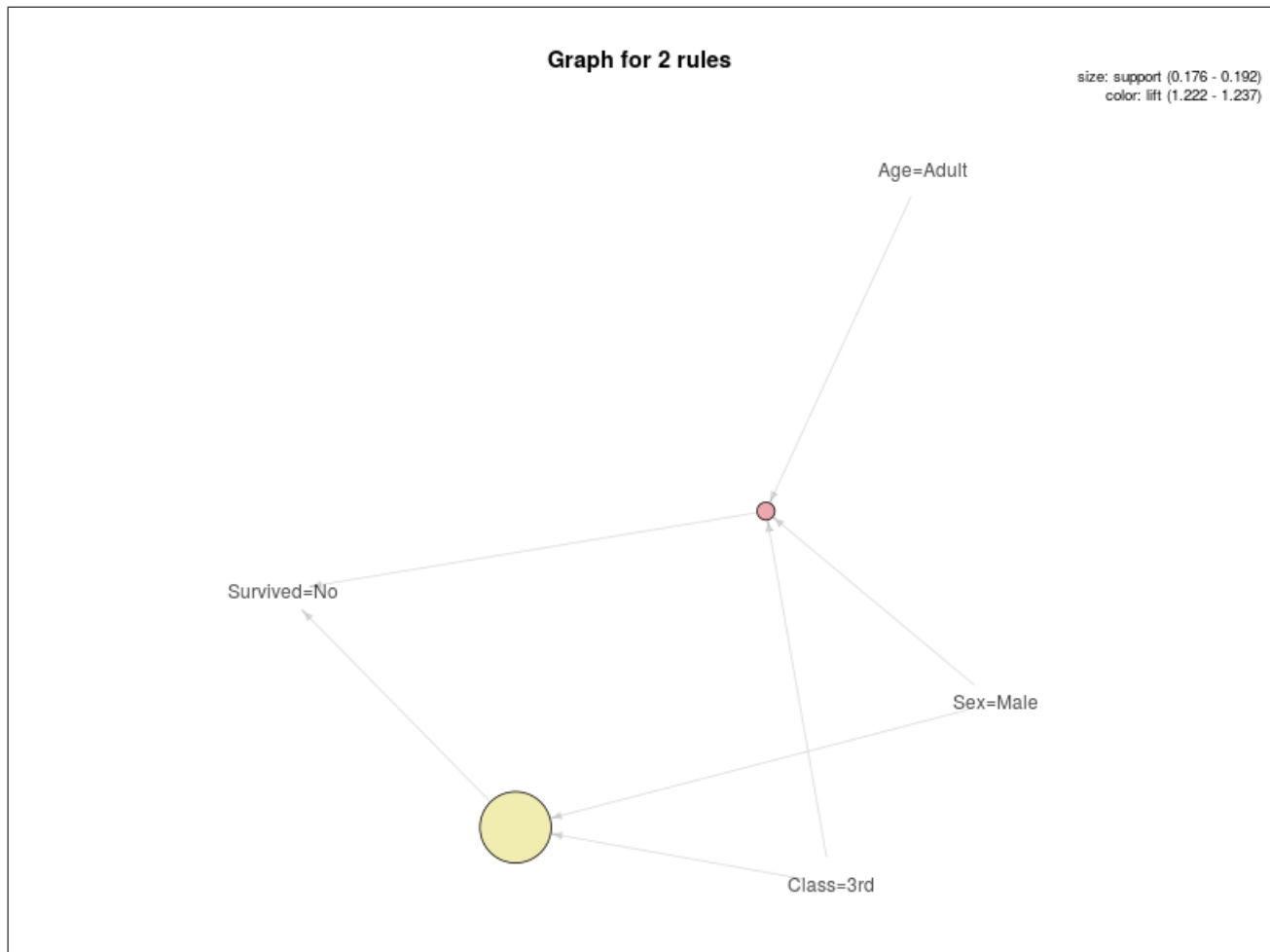
For this question I modified the previous question code to increase the support to 0.1 here is the code :


```

1 rules = apriori(titanic,parameter = list(minlen=2, supp=0.1, conf=0.8),appearance = list(rhs=c
  ("Survived=No", "Survived=Yes"),default="lhs"))
2 inspect(rules)
3 #visualization
4 library(arulesViz)
5 png('rules.png',width = 800,height = 600)
6 plot(rules, method="graph", control=list(type="items"))
7 dev.off()

```

The resulted figure :



The rules are :

(Class=3rd,Sex=Male=> Survived =NO)

(Class=3rd,Age=Adult,Sex=Male=> Survived =NO)

The only thing I changed was the support to be 0.1 instead of 0.05.

Sixth Question

For this question I plotted 3 measurements.Each measurement is plotted against four elements (F_{00} , F_{01} , F_{11} , F_{10}) and here is the code I used to plot all these measurements:

```

1 def doregularthings(xlab,ylab,title,ymin,ymax):
2     plt.xlim(0,10000)
3     #plt.ylim(ymin,ymax)
4     plt.ylim(-30,0)
5     plt.title(title)
6     plt.ylabel(ylab)
7     plt.xlabel(xlab)
8     plt.figure(1)
9     ###Laplace#####
10    #F11
11    plt.subplot(221)

```

```

12 plt.scatter([x.f11 for x in lst],[x.laplace for x in lst])
13 doregularthings('F11','laplace','Laplace vs F11',0,1)
14
15 #F10
16 plt.subplot(222)
17 plt.scatter([x.f10 for x in lst],[x.laplace for x in lst])
18 doregularthings('F10','laplace','Laplace vs F10',0,1)
19
20 #F01
21 plt.subplot(223)
22 plt.scatter([x.f01 for x in lst],[x.laplace for x in lst])
23 doregularthings('F01','laplace','Laplace vs F01',0,1)
24 #F00
25 plt.subplot(224)
26 plt.scatter([x.f00 for x in lst],[x.laplace for x in lst])
27 doregularthings('F00','laplace','Laplace vs F00',0,1)
28 plt.show()
29
30 plt.figure(2)
31 ####conviction#####
32 #F11
33 plt.subplot(221)
34 plt.scatter([x.f11 for x in lst],[x.conviction for x in lst])
35 doregularthings('F11','conviction','conviction vs F11',0,20)
36
37 #F10
38 plt.subplot(222)
39 plt.scatter([x.f10 for x in lst],[x.conviction for x in lst])
40 doregularthings('F10','conviction','conviction vs F10',0,20)
41
42 #F01
43 plt.subplot(223)
44 plt.scatter([x.f01 for x in lst],[x.conviction for x in lst])
45 doregularthings('F01','conviction','conviction vs F01',0,20)
46 #F00
47 plt.subplot(224)
48 plt.scatter([x.f00 for x in lst],[x.conviction for x in lst])
49 doregularthings('F00','conviction','conviction vs F00',0,20)
50 plt.show()
51
52 plt.figure(3)
53 ####Certain Factor#####
54 #F11
55 plt.subplot(221)
56 plt.scatter([x.f11 for x in lst],[x.certinfactor for x in lst])
57 doregularthings('F11','Certain Factor','Certain Factor vs F11',0,1)
58
59 #F10
60 plt.subplot(222)
61 plt.scatter([x.f10 for x in lst],[x.certinfactor for x in lst])
62 doregularthings('F10','Certain Factor','Certain Factor vs F10',0,1)
63
64 #F01
65 plt.subplot(223)
66 plt.scatter([x.f01 for x in lst],[x.certinfactor for x in lst])
67 doregularthings('F01','Certain Factor','Certain Factor vs F01',0,1)
68 #F00
69 plt.subplot(224)
70 plt.scatter([x.f00 for x in lst],[x.certinfactor for x in lst])
71 doregularthings('F00','Certain Factor','Certain Factor vs F00',0,1)
72 plt.show()

```

Here is the figure generated by the previous code :

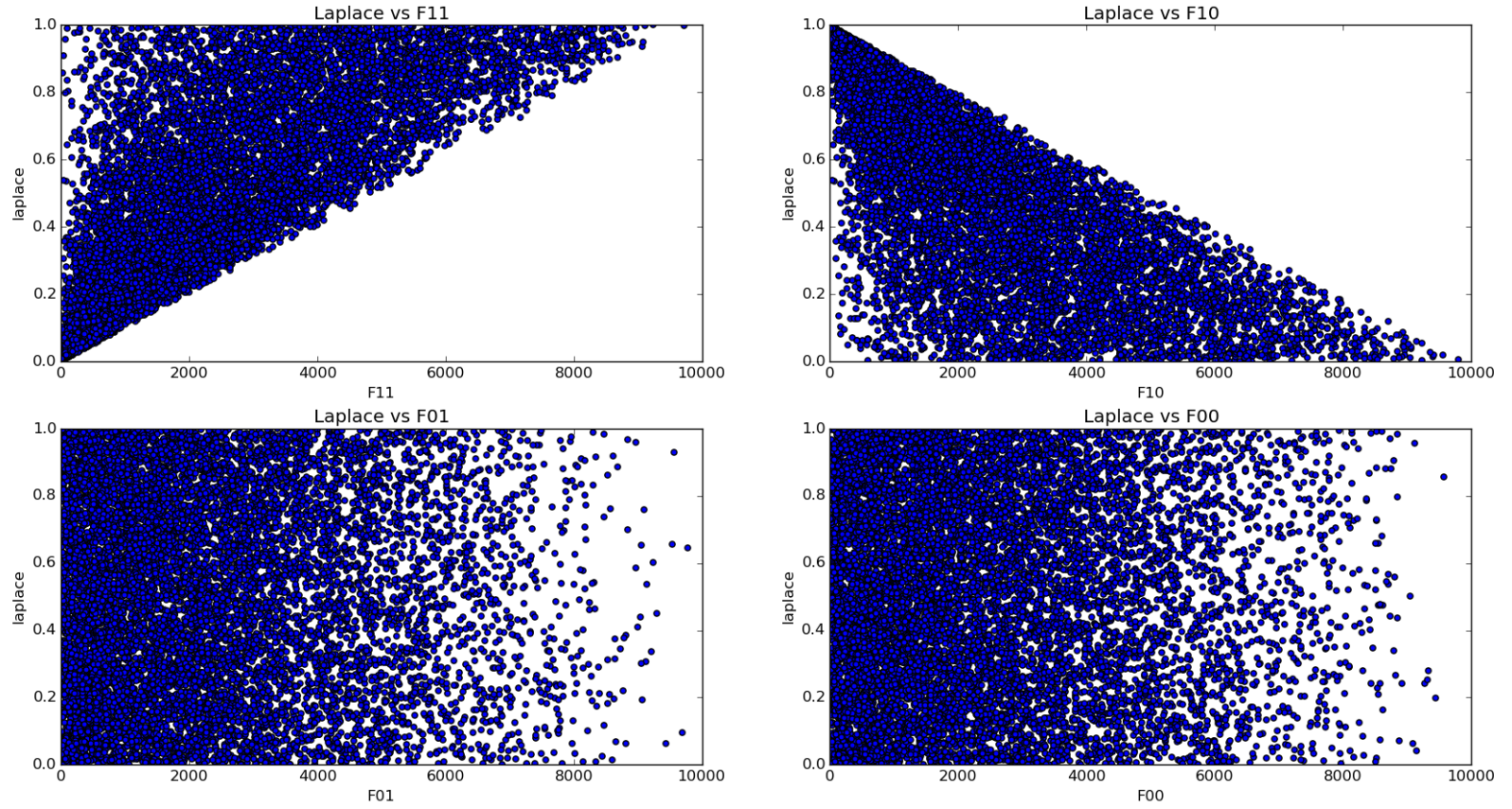


Figure 7: Laplace vs confusion matrix

From the figures I couldn't get any relation between Laplace and F_{01}, F_{00} but I can say that F_{11} start limit for Laplace measurement. While F_{10} end limit.

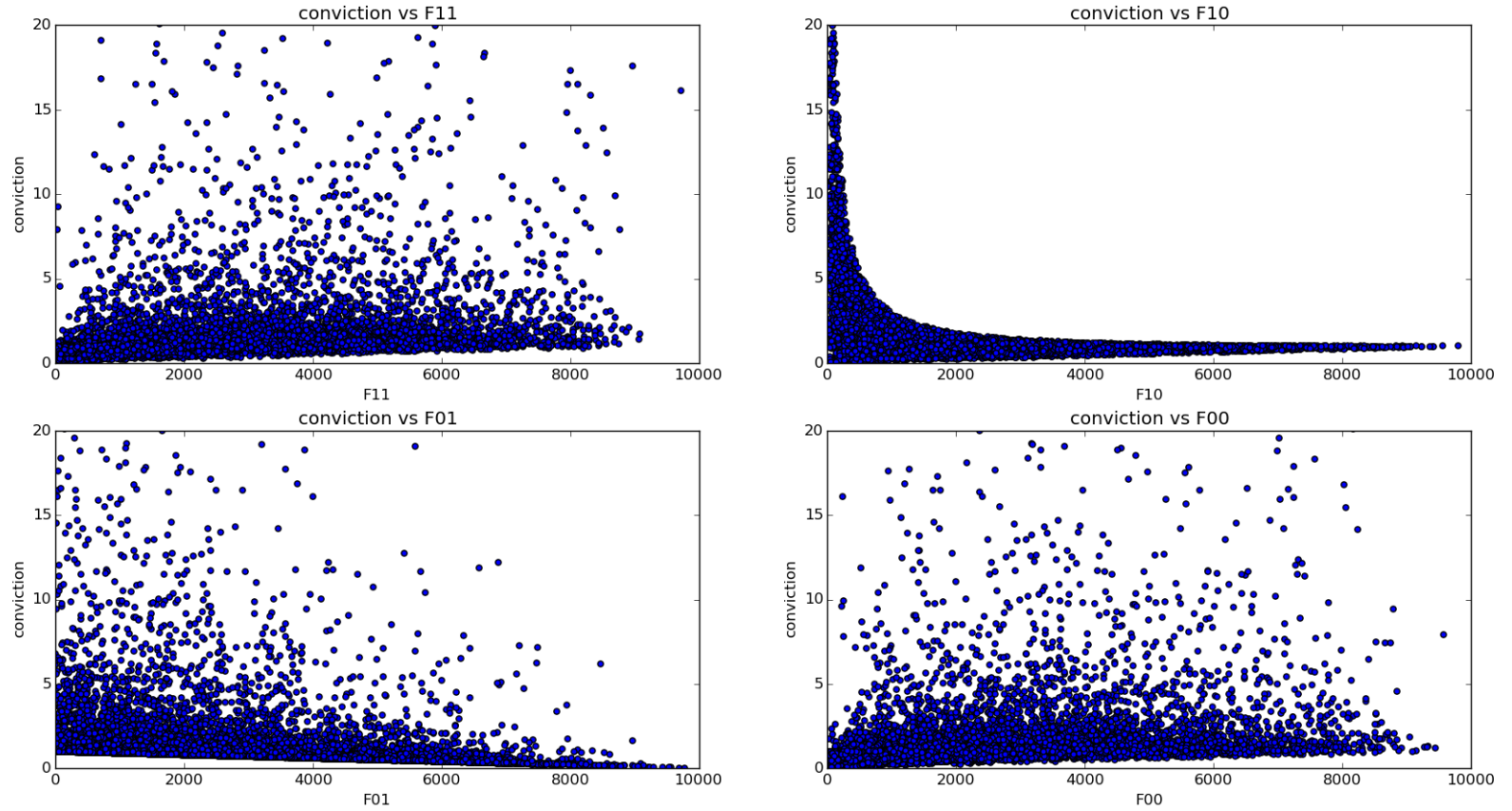


Figure 8: Conviction vs confusion matrix

from In this figure I modified *ylim* to get the value focused on the area with main density. In figure 8 we notice a pattern with F_{10} . it looks like Hyperbola on the domain $]0, +\infty[$ something like $\frac{1}{x}$

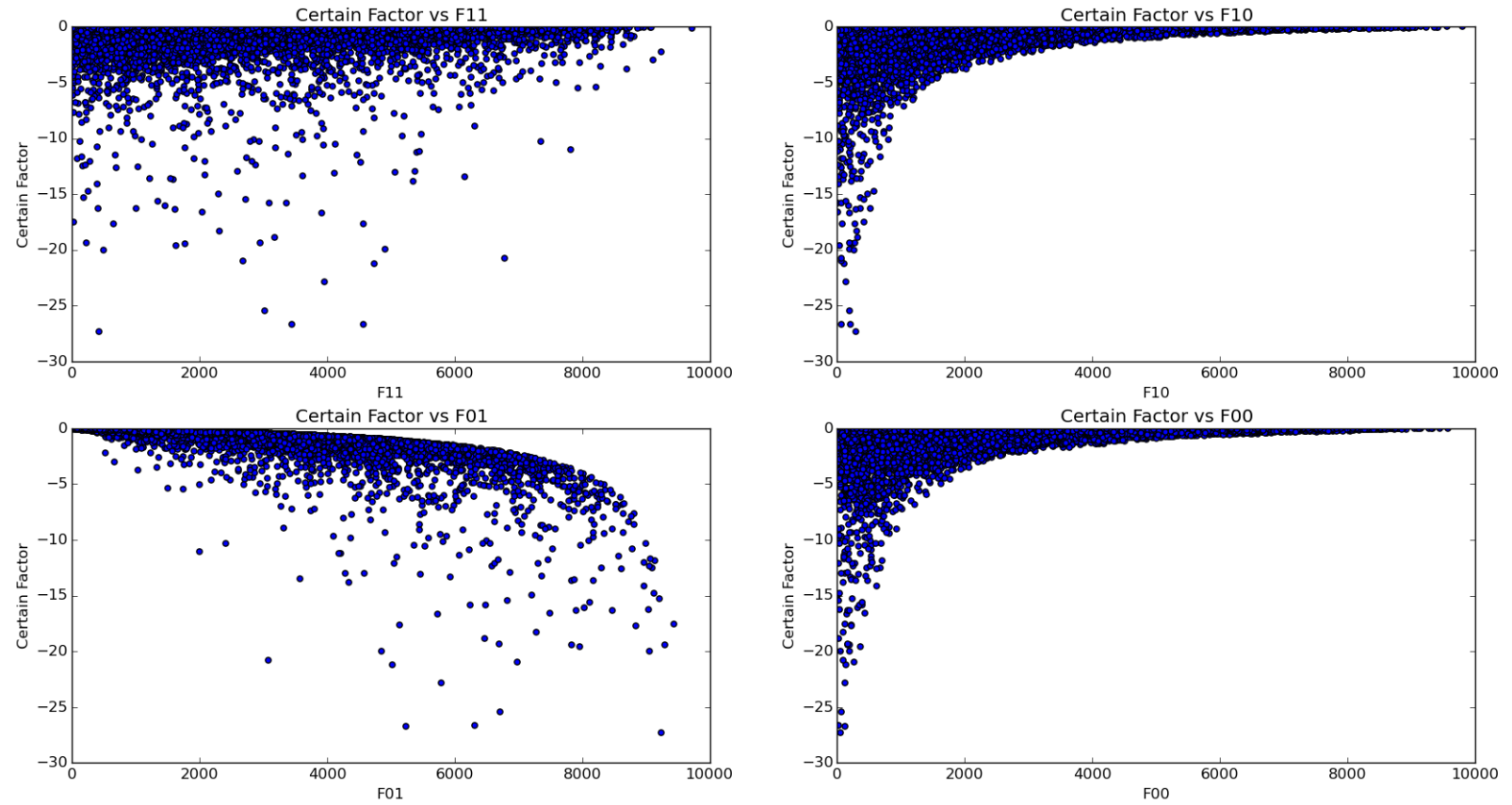


Figure 9: Certainty factor vs confusion matrix

In figure 9 I think certainty factor is related to F_{10}, F_{00} and it looks like $\frac{-1}{x}$

Note: All code, tex, ipython, R, py, png files exist on github

E.O.F