

# titanic

April 17, 2024

## 0.1 Titanic with Apriori

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[23]: import pandas as pd # type: ignore
import numpy as np # type: ignore
from mlxtend.frequent_patterns import apriori, association_rules
import matplotlib.pyplot as plt

df = pd.read_csv('titanic.csv')
df = df.drop(df.columns[0], axis=1)

data = pd.get_dummies(df, columns=['Class', 'Sex', 'Age', 'Survived'])

#print(data.head())

freq = apriori(data, min_support=0.005, use_colnames=True, verbose=1)

#print(freq.head(20))

rules = association_rules(freq, metric="confidence", min_threshold=0.8)
rules = rules.sort_values('confidence', ascending=False)
print(rules)

# Filtrowanie reguł dotyczących przeżywalności
survival_rules = rules[rules['consequents'].apply(lambda x: ('Survived_Yes' in
↳str(x) or 'Survived_No' in str(x)) and 'Age_Adult' not in str(x))]

# Sortowanie reguł wg ufności
survival_rules = survival_rules.sort_values('confidence', ascending=False)

# Wyświetlenie najciekawszych reguł
print(survival_rules)

# Utworzenie wykresu słupkowego
plt.figure(figsize=(12, 6))
plt.barh(range(len(survival_rules)), survival_rules['confidence'],
↳color='skyblue')
plt.yticks(range(len(survival_rules)), survival_rules['antecedents'].
↳apply(lambda x: ', '.join(list(x))))
```

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plt.xlabel('Confidence')
plt.title('Survival')
plt.gca().invert_yaxis() # Odwróć oś Y, aby najważniejsze reguły były na górze
plt.show()
```

Processing 40 combinations | Sampling itemset size 54

	antecedents	consequents	antecedent support \
65	(Survived_No, Sex_Female, Age_Child)	(Class_3rd)	0.007724
22	(Class_2nd, Age_Child)	(Survived_Yes)	0.010904
57	(Sex_Male, Survived_No, Class_2nd)	(Age_Adult)	0.069968
70	(Sex_Male, Survived_No, Age_Child)	(Class_3rd)	0.015902
56	(Sex_Female, Class_2nd, Age_Child)	(Survived_Yes)	0.005906
..	...	...	...
62	(Class_3rd, Survived_No, Sex_Female)	(Age_Adult)	0.048160
67	(Class_3rd, Age_Adult, Sex_Male)	(Survived_No)	0.209905
26	(Class_3rd, Sex_Male)	(Survived_No)	0.231713
63	(Survived_No, Age_Adult, Sex_Female)	(Class_3rd)	0.049523
66	(Class_3rd, Survived_No, Age_Adult)	(Sex_Male)	0.216265

	consequent support	support	confidence	lift	leverage	conviction \
65	0.320763	0.007724	1.000000	3.117564	0.005246	inf
22	0.323035	0.010904	1.000000	3.095640	0.007382	inf
57	0.950477	0.069968	1.000000	1.052103	0.003465	inf
70	0.320763	0.015902	1.000000	3.117564	0.010801	inf
56	0.323035	0.005906	1.000000	3.095640	0.003998	inf
..	...	...	...	...	...	...
62	0.950477	0.040436	0.839623	0.883370	-0.005339	0.308790
67	0.676965	0.175829	0.837662	1.237379	0.033731	1.989896
26	0.676965	0.191731	0.827451	1.222295	0.034870	1.872135
63	0.320763	0.040436	0.816514	2.545534	0.024551	3.701840
66	0.786461	0.175829	0.813025	1.033777	0.005745	1.142075

	zhangs_metric
65	0.684524
22	0.684428
57	0.053249
70	0.690212
56	0.680987
..	...
62	-0.121813
67	0.242806
26	0.236717
63	0.638790
66	0.041690

[80 rows x 10 columns]

antecedents	consequents	antecedent support \
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22	(Class_2nd, Age_Child)	(Survived_Yes)	0.010904
56	(Sex_Female, Class_2nd, Age_Child)	(Survived_Yes)	0.005906
11	(Class_1st, Sex_Female)	(Survived_Yes)	0.065879
46	(Class_1st, Age_Adult, Sex_Female)	(Survived_Yes)	0.065425
58	(Sex_Male, Age_Adult, Class_2nd)	(Survived_No)	0.076329
17	(Sex_Female, Class_2nd)	(Survived_Yes)	0.048160
31	(Sex_Female, Class_Crew)	(Survived_Yes)	0.010450
72	(Age_Adult, Class_Crew, Sex_Female)	(Survived_Yes)	0.010450
19	(Sex_Male, Class_2nd)	(Survived_No)	0.081327
55	(Class_2nd, Age_Adult, Sex_Female)	(Survived_Yes)	0.042254
67	(Class_3rd, Age_Adult, Sex_Male)	(Survived_No)	0.209905
26	(Class_3rd, Sex_Male)	(Survived_No)	0.231713

	consequent	support	support	confidence	lift	leverage	conviction	\
22	0.323035	0.010904	1.000000	3.095640	0.007382	inf		
56	0.323035	0.005906	1.000000	3.095640	0.003998	inf		
11	0.323035	0.064062	0.972414	3.010243	0.042781	24.539982		
46	0.323035	0.063607	0.972222	3.009650	0.042473	24.370741		
58	0.676965	0.069968	0.916667	1.354083	0.018296	3.876420		
17	0.323035	0.042254	0.877358	2.715986	0.026696	5.519869		
31	0.323035	0.009087	0.869565	2.691861	0.005711	5.190065		
72	0.323035	0.009087	0.869565	2.691861	0.005711	5.190065		
19	0.676965	0.069968	0.860335	1.270871	0.014913	2.312930		
55	0.323035	0.036347	0.860215	2.662916	0.022698	4.842904		
67	0.676965	0.175829	0.837662	1.237379	0.033731	1.989896		
26	0.676965	0.191731	0.827451	1.222295	0.034870	1.872135		

	zhangs_metric
22	0.684428
56	0.680987
11	0.714898
46	0.714480
58	0.283102
17	0.663777
31	0.635147
72	0.635147
19	0.232006
55	0.652022
67	0.242806
26	0.236717

