

Association Rules Mining

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Statistics for Data Science
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Dataset:-

An individual's annual income results from various factors.

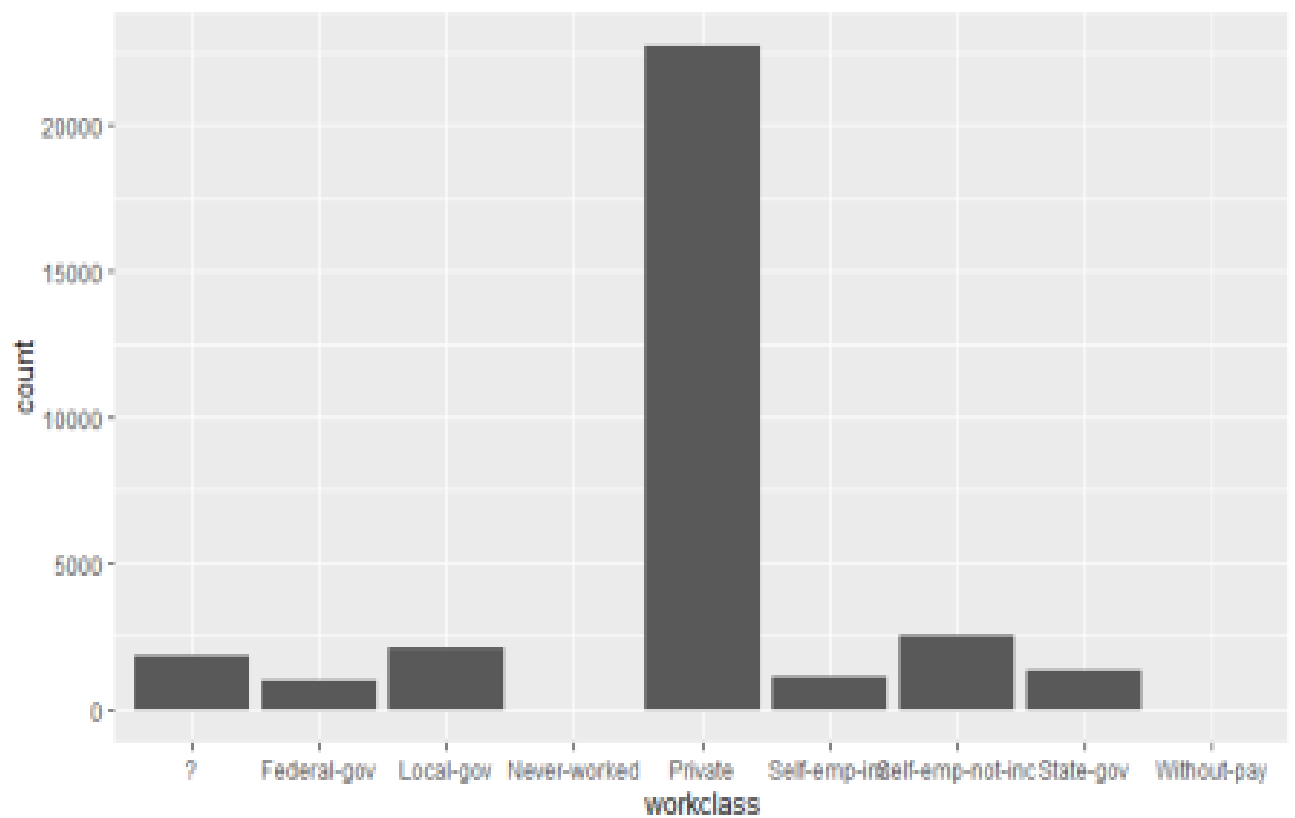
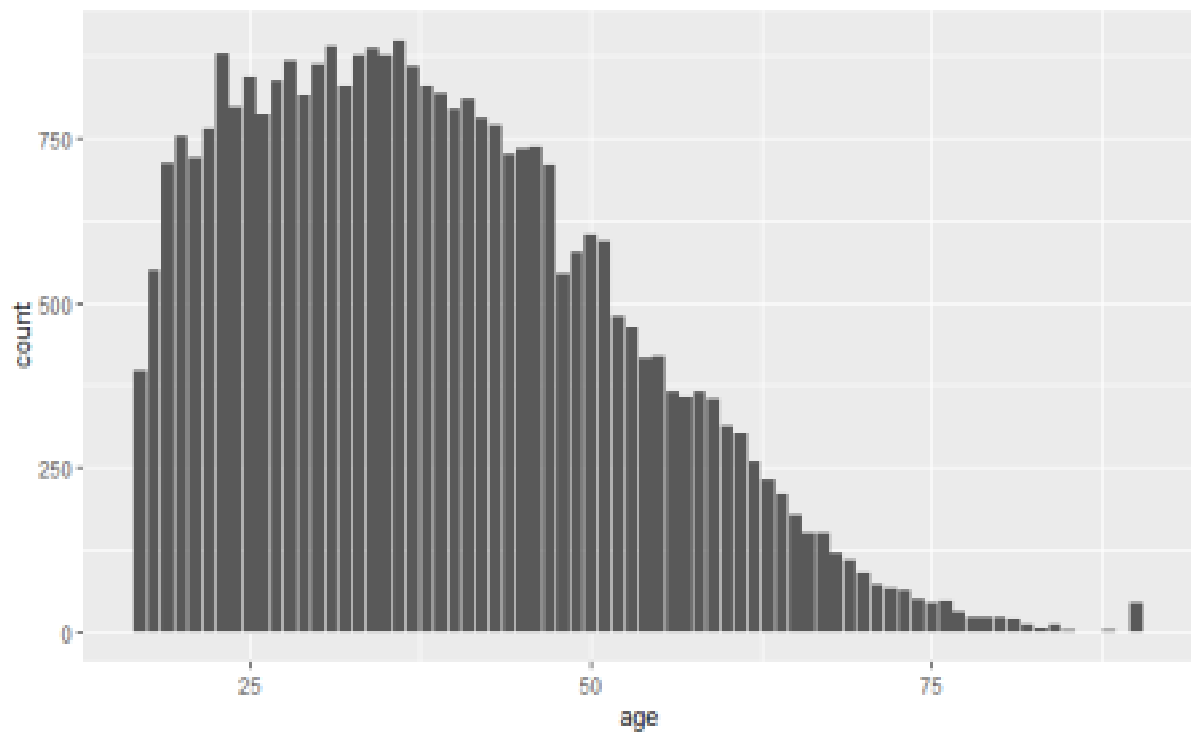
Income is influenced by the individual's education level, age, gender, occupation, etc.

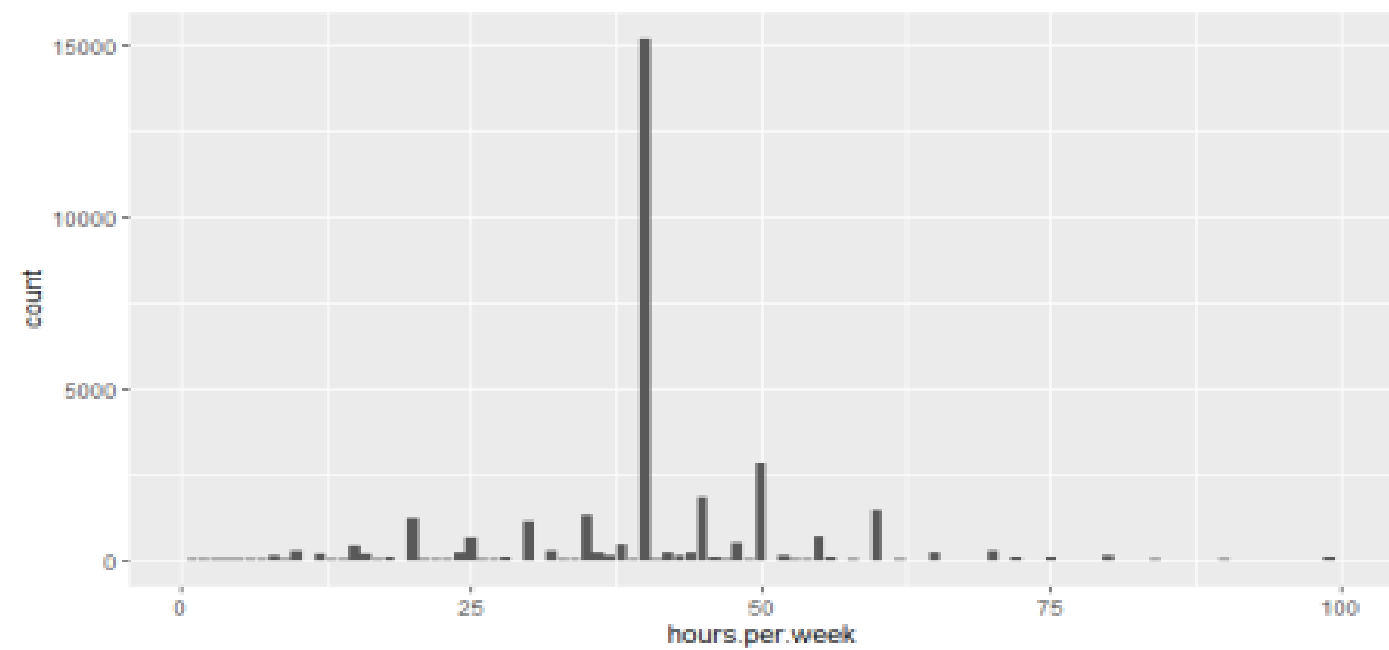
Fields: The dataset contains 16 columns.

Target Field: Income

The income is divided into two classes: $\leq 50K$ and $> 50K$.

Data Visualization:-





Data Discretization:

Here we discretized 4 columns: age, capital.gain, capital.loss, hours.per.week

Terms Used:

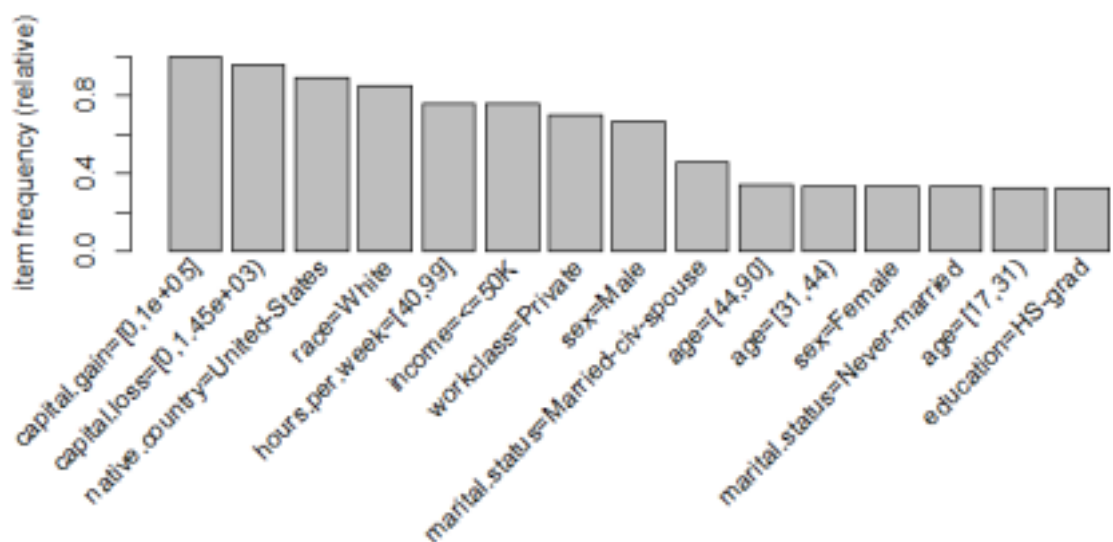
Lift → Probability of togetherness

Gini → Info Provide

Phi → Binary Correlation

Support → frequency of items occurred.

Confidence → How Reliable the rule is.



Maximal frequent itemset:

Maximal frequent itemset shows that none of its immediate supersets are frequent.

Here, the total no. of maximal frequent itemsets is 8. Some of them are:

Items	Support	Count
{race=White, capital.gain=[0,1e+05], capital.loss=[0,1.45e+03), hours.per.week=[40,99], native.country=United States}	0.5709	18592

{race=White, capital.gain=[0,1e+05], capital.loss=[0,1.45e+03), native.country=United States, income=<=50K}	0.5645	18382
{workclass=Private, capital.gain=[0,1e+05], capital.loss=[0,1.45e+03), income=<=50K}	0.5307	17281
{capital.gain=[0,1e+05],	0.5279	17192

capital.loss=[0,1.45e+03), hours.per.week=[40,99], income=<=50K}		
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Closest frequent itemset:

Closest frequent itemset refers that none of its immediate superset has same support.

Here are total 48 itemsets of such type, some of them are:

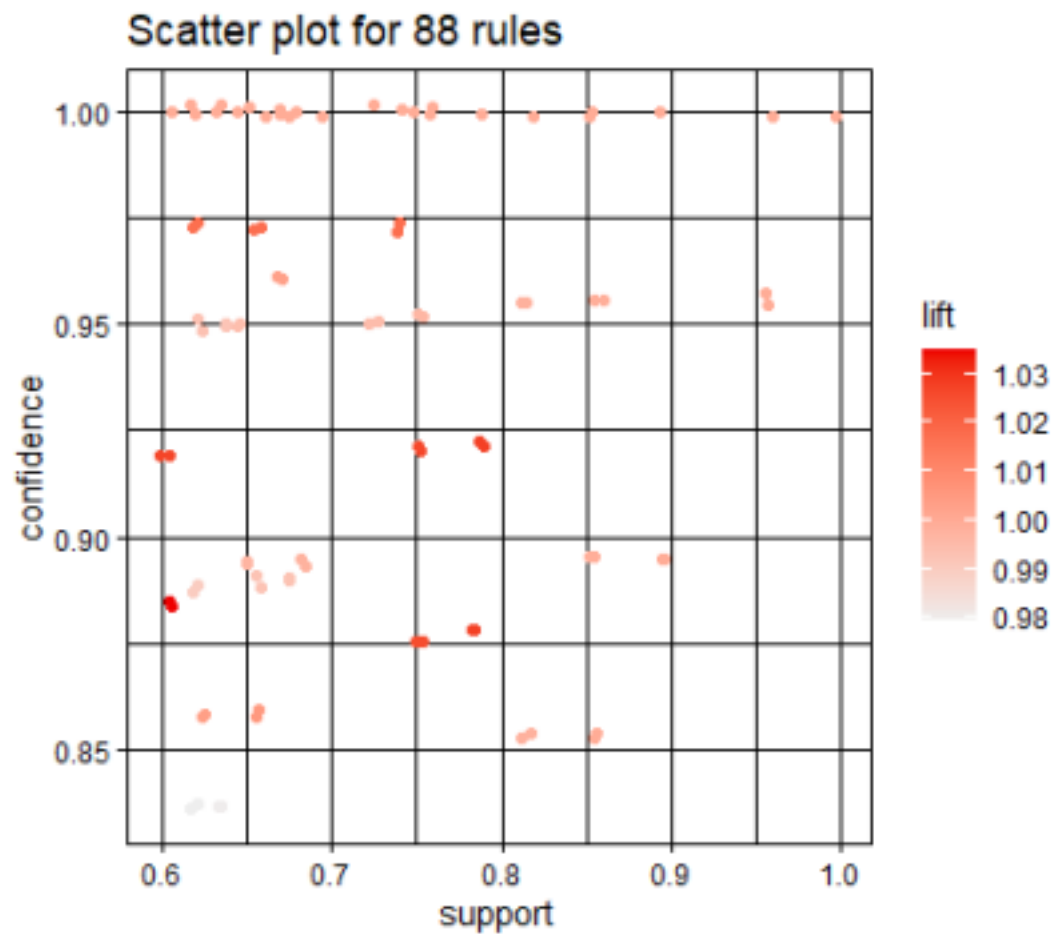
Item	Support	Count
{capital.gain=[0,1e+05]}	1.000	32561
{capital.gain=[0,1e+05], capital.loss=[0,1.45e+03)}	0.9559	31127

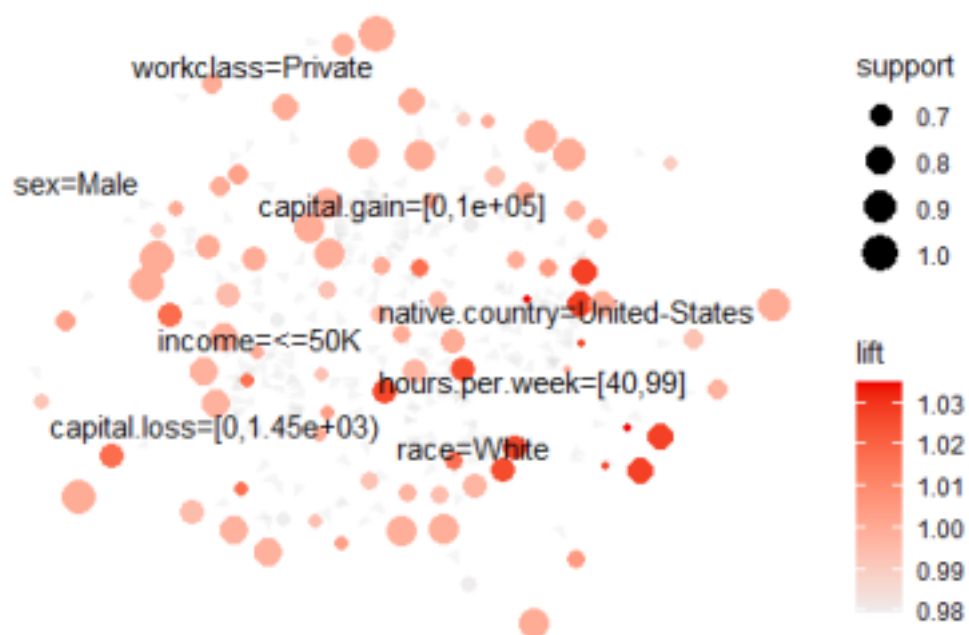
Association rule:

support = 0.6

confidence = 0.8

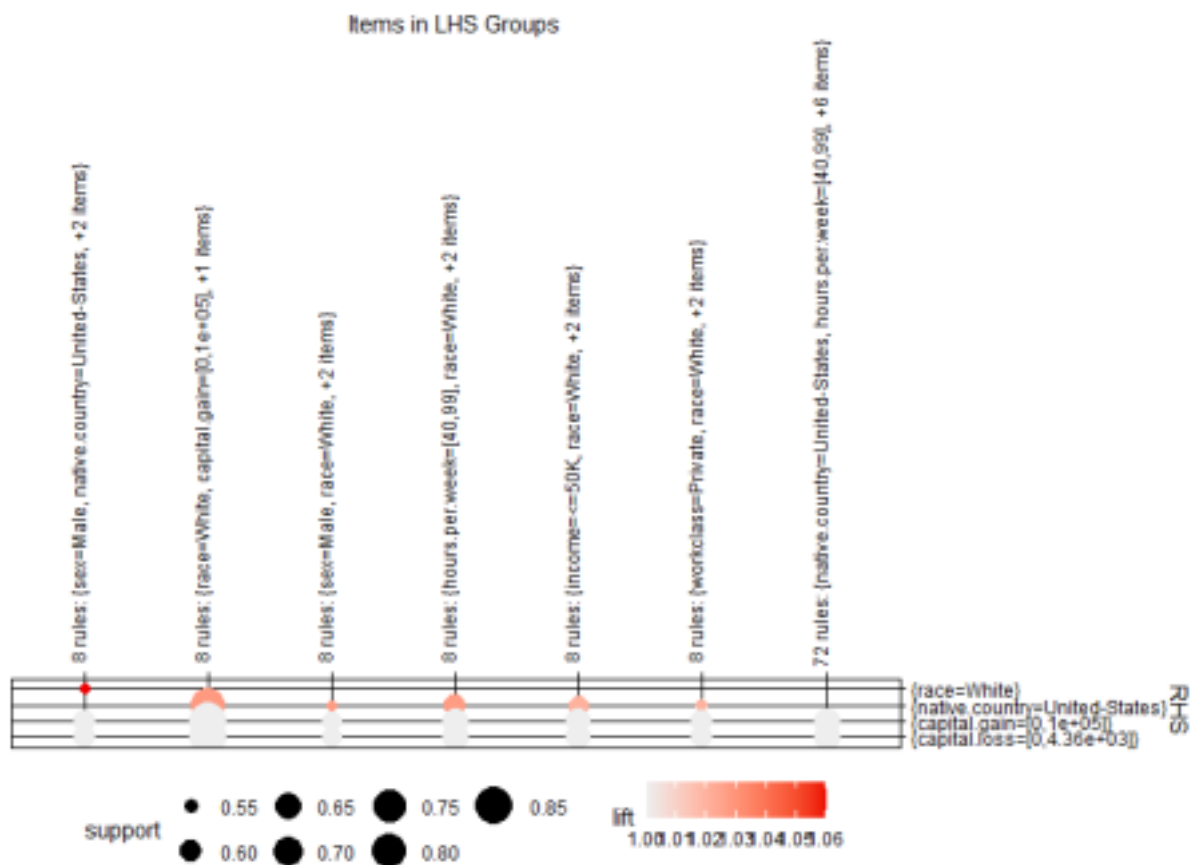
Total number of association rules is 88.





Bigger the circle, higher the support.

Higher the intensity, higher the lift(probability).



	lhs	rhs	support	confidence	coverage	lift	count
[1]	{hours.per.week=[40,99], native.country=United-States}	=> {race=white}	0.6021314	0.8841887	0.6809987	1.035018	19606
[2]	{capital.gain=[0,1e+05], hours.per.week=[40,99], native.country=United-States}	=> {race=white}	0.6021314	0.8841887	0.6809987	1.035018	19606
[3]	{race=white}	=> {native.country=United-States}	0.7868616	0.9210886	0.8542735	1.028165	25621
[4]	{race=white, capital.gain=[0,1e+05]}	=> {native.country=United-States}	0.7868616	0.9210886	0.8542735	1.028165	25621
[5]	{native.country=United-States}	=> {race=white}	0.7868616	0.8783339	0.8958570	1.028165	25621
[6]	{capital.gain=[0,1e+05], native.country=United-States}	=> {race=white}	0.7868616	0.8783339	0.8958570	1.028165	25621
[7]	{race=white, hours.per.week=[40,99]}	=> {native.country=United-States}	0.6021314	0.9203399	0.6542489	1.027329	19606
[8]	{race=white, capital.gain=[0,1e+05], hours.per.week=[40,99]}	=> {native.country=United-States}	0.6021314	0.9203399	0.6542489	1.027329	19606
[9]	{race=white, capital.loss=[0,1.45e+03]}	=> {native.country=United-States}	0.7499770	0.9198780	0.8153005	1.026813	24420
[10]	{race=white, capital.gain=[0,1e+05], capital.loss=[0,1.45e+03]}	=> {native.country=United-States}	0.7499770	0.9198780	0.8153005	1.026813	24420
[11]	{capital.loss=[0,1.45e+03], native.country=United-States}	=> {race=white}	0.7499770	0.8762739	0.8558705	1.025753	24420
[12]	{capital.gain=[0,1e+05], capital.loss=[0,1.45e+03], native.country=United-States}	=> {race=white}	0.7499770	0.8762739	0.8558705	1.025753	24420
[13]	{income<=50K}	=> {capital.loss=[0,1.45e+03]}	0.7388287	0.9731796	0.7591904	1.018013	24057
[14]	{capital.gain=[0,1e+05], income<=50K}	=> {capital.loss=[0,1.45e+03]}	0.7388287	0.9731796	0.7591904	1.018013	24057
[15]	{native.country=United-States, income<=50K}	=> {capital.loss=[0,1.45e+03]}	0.6572280	0.9727715	0.6756242	1.017586	21400