

A-Priori Algorithm

Association Rule Learning Apriori Intuition

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ARL - What is it all about ?



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ARL - What is it all about ?

People who bought also bought ...

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ARL - Movie Recommendation

User ID	Movies liked
46578	Movie1, Movie2, Movie3, Movie4
98989	Movie1, Movie2
71527	Movie1, Movie2, Movie4
78981	Movie1, Movie2
89192	Movie2, Movie4
61557	Movie1, Movie3

Potential Rules:

Movie1 \Rightarrow Movie2

Movie2 \Rightarrow Movie4

Movie1 \Rightarrow Movie3

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ARL - Market Basket Optimisation

Transaction ID	Products purchased
46578	Burgers, French Fries, Vegetables
98989	Burgers, French Fries, Ketchup
71527	Vegetables, Fruits
78981	Pasta, Fruits, Butter, Vegetables
89192	Burgers, Pasta, French Fries
61557	Fruits, Orange Juice, Vegetables
87923	Burgers, French Fries, Ketchup, Mayo

Potential Rules:

Burgers \Rightarrow French Fries

Vegetables \Rightarrow Fruits

Burgers, French Fries \Rightarrow Ketchup

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Apriori - Support

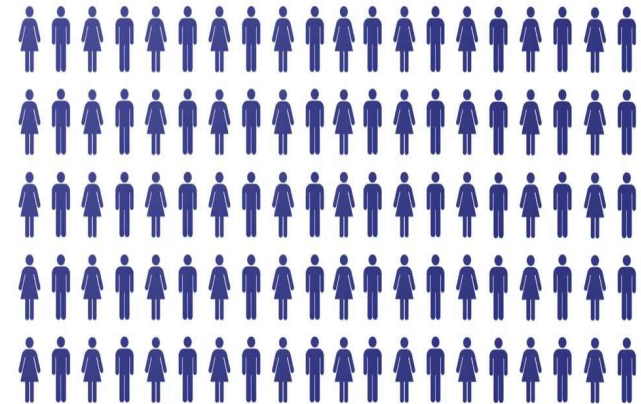
Movie Recommendation: $\text{support}(M) = \frac{\# \text{ user watchlists containing } M}{\# \text{ user watchlists}}$

Market Basket Optimisation: $\text{support}(I) = \frac{\# \text{ transactions containing } I}{\# \text{ transactions}}$

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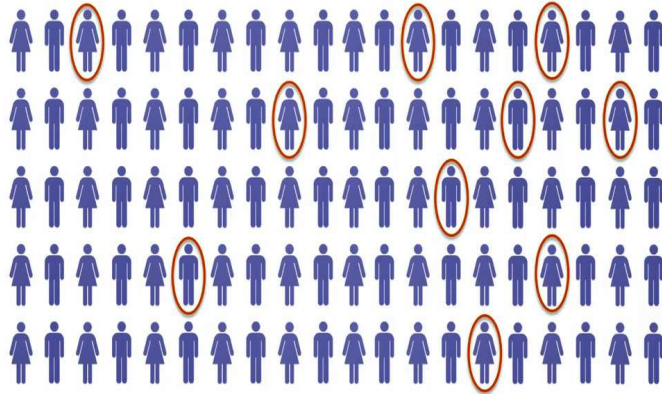
Apriori - Support



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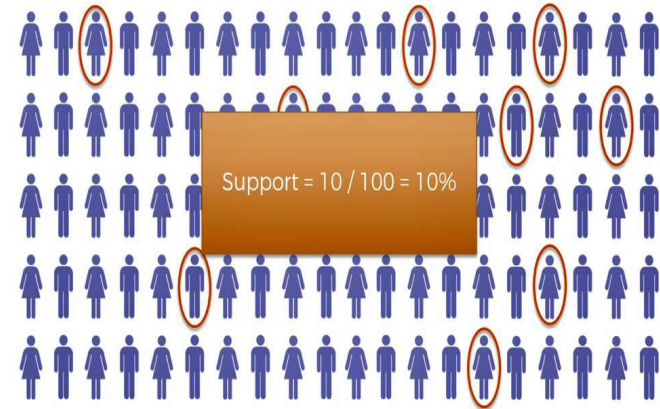
Apriori - Support



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Apriori - Support



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Apriori - Confidence

Movie Recommendation: $\text{confidence}(M_1 \rightarrow M_2) = \frac{\# \text{ user watchlists containing } M_1 \text{ and } M_2}{\# \text{ user watchlists containing } M_1}$

Market Basket Optimisation: $\text{confidence}(I_1 \rightarrow I_2) = \frac{\# \text{ transactions containing } I_1 \text{ and } I_2}{\# \text{ transactions containing } I_1}$

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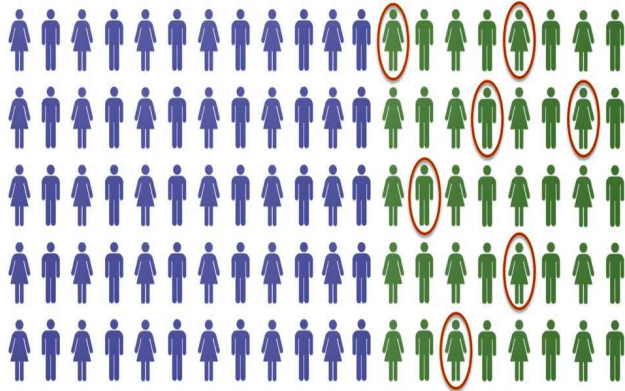
Apriori - Confidence



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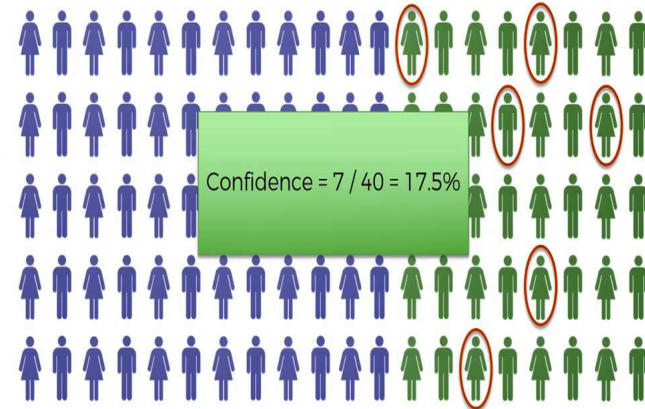
Apriori - Confidence



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Apriori - Confidence



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Apriori - Lift

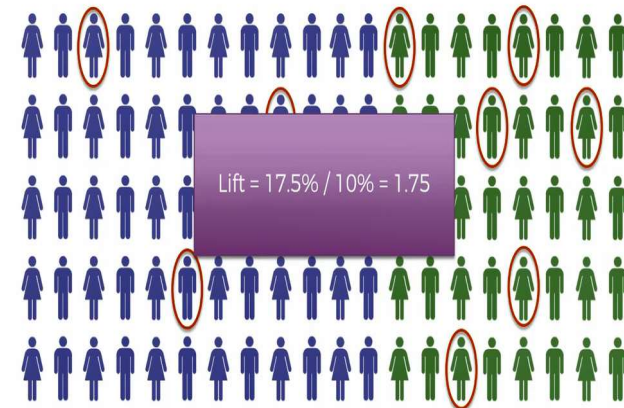
Movie Recommendation:
$$\text{lift}(M_1 \rightarrow M_2) = \frac{\text{confidence}(M_1 \rightarrow M_2)}{\text{support}(M_2)}$$

Market Basket Optimisation:
$$\text{lift}(I_1 \rightarrow I_2) = \frac{\text{confidence}(I_1 \rightarrow I_2)}{\text{support}(I_2)}$$

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Apriori - Lift



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Apriori - Algorithm

Step 1: Set a minimum support and confidence



Step 2: Take all the subsets in transactions having higher support than minimum support



Step 3: Take all the rules of these subsets having higher confidence than minimum confidence



Step 4: Sort the rules by decreasing lift

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- The Apriori Algorithm is an algorithm for mining frequent itemsets for Boolean association rules.
- Apriori uses a "bottom up" approach, where frequent subsets are extended one item at a time which is known as candidate generation, and groups of candidates are tested against the data.

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Example

Design the association rules for the given table using Apriori Algorithm.

Transaction ID	Itemsets
1	A,B,C
2	A,C
3	A,D
4	B,E,F

Support = 2
Confidence = 50%

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Candidate C1

Transaction ID	Itemsets
1	A,B,C
2	A,C
3	A,D
4	B,E,F

Itemset	Support
{A}	3
{B}	2
{C}	2
{D}	1
{E}	1
{F}	1

Support = 2
Confidence = 50%

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Candidate C1

Transaction ID	Itemsets
1	A,B,C
2	A,C
3	A,D
4	B,E,F

Itemset	Support
{A}	3
{B}	2
{C}	2
{D}	1
{E}	1
{F}	1

Support = 2
Confidence = 50%

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Candidate C1

Transaction ID	Itemsets
1	A,B,C
2	A,C
3	A,D
4	B,E,F

L1 [Frequent Itemset]

Itemset	Support
{A}	3
{B}	2
{C}	2

Support = 2
Confidence = 50%

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Candidate C2

Transaction ID	Itemsets
1	A,B,C
2	A,C
3	A,D
4	B,E,F

Itemset	Support
{A}	3
{B}	2
{C}	2

Itemset	Support
{A,B}	1
{B,C}	1
{A,C}	2

Support = 2
Confidence = 50%

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Candidate C2

Transaction ID	Itemsets
1	A,B,C
2	A,C
3	A,D
4	B,E,F

Itemset	Support
{A}	3
{B}	2
{C}	2

Itemset	Support
{A,B}	1
{B,C}	1
{A,C}	2

Support = 2
Confidence = 50%

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Candidate C2

Transaction ID	Itemsets
1	A,B,C
2	A,C
3	A,D
4	B,E,F

Itemset	Support
{A}	3
{B}	2
{C}	2

L2 [Frequent Itemset]

Itemset	Support
{A,C}	2

Support = 2
Confidence = 50%

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Candidate C2

Itemset	Support
{A,C}	2

Associative Rule	Support	Confidence	Confidence %
A -> C	2	$2/3 = 0.66$	66%
C -> A	2	$2/2 = 1$	100%

Support = 2
Confidence = 50%



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Example 2

- A database has 5 transactions
- Minimum support = 60%
- Minimum confidence = 80%

T_ID	Itemsets
T_1000	M,O,N,K,E,Y
T_1001	D,O,N,K,E,Y
T_1002	M,A,K,E
T_1003	M,U,C,K,Y
T_1004	C,O,O,K,E

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Candidate C1

Itemset	Support
{M}	3
{O}	3
{N}	2
{K}	5
{E}	4
{Y}	3
{D}	1
{A}	1
{U}	1
{C}	2

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■ L1

Itemset	Support
{M}	3
{O}	3
{N}	2
{K}	5
{E}	4
{Y}	3
{D}	1
{A}	1
{U}	1
{C}	2

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■ L1

Itemset	Support
{M}	3
{O}	3
{K}	5
{E}	4
{Y}	3

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■ Candidate C2

Itemset	Support
{M,O}	1
{M,K}	3
{M,E}	2
{M,Y}	2
{O,K}	3
{O,E}	3
{O,Y}	2
{K,E}	4
{K,Y}	3
{E,Y}	2

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■ L2

Itemset	Support
{M,O}	1
{M,K}	3
{M,E}	2
{M,Y}	2
{O,K}	3
{O,E}	3
{O,Y}	2
{K,E}	4
{K,Y}	3
{E,Y}	2

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■ L₂

Itemset	Support
{M,K}	3
{O,K}	3
{O,E}	3
{K,E}	4
{K,Y}	3

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■ Candidate C₃

Itemset	Support
{M,O,K}	1
{M,K,O,E}	1
{M,K,E}	2
{M,K,Y}	2
{O,K,E}	3
{O,K,Y}	2
{O,K,E,Y}	2
{K,E,Y}	2

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■ L₃

Itemset	Support
{M,O,K}	1
{M,K,O,E}	1
{M,K,E}	2
{M,K,Y}	2
{O,K,E}	3
{O,K,Y}	2
{O,K,E,Y}	2
{K,E,Y}	2

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■ L₃

Itemset	Support
{O,K,E}	3

■ Association rules

Association Rules	Support	Confidence	Confidence %
$O \rightarrow K^{\wedge}E$	3	$3/3 = 1$	100
$K \rightarrow O^{\wedge}E$	3	$3/5 = 0.6$	60
$E \rightarrow K^{\wedge}O$	3	$3/4 = 0.75$	75

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■ Association rules

The minimum confidence is 80%

Association Rules	Support	Confidence	Confidence %
$O \rightarrow K^E$	3	$3/3 = 1$	100
$K \rightarrow O^E$	3	$3/5 = 0.6$	60
$E \rightarrow K^O$	3	$3/4 = 0.75$	75

$O \rightarrow K^E$ = ACCEPTED

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Design association rules for a database has 9 transactions

Minimum support = 2

Minimum confidence = 70% ✓

ID	Itemsets
100	1,2,5
200	2,4
300	2,3
400	1,2,4
500	1,3
600	1,3
700	1,3,2,5
800	1,3
900	1,2,3

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■ Candidate C1

Itemset	Support
{1}	7
{2}	6
{3}	6
{4}	2
{5}	2

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■ Candidate C2

Itemset	Support
{1,2}	3
{1,3}	5
{1,4}	1
{1,5}	2
{2,3}	3
{2,4}	2
{2,5}	2
{3,4}	-
{3,5}	1
{4,5}	-


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■ L2

Itemset	Support
{1,2}	3
{1,3}	5
{1,4}	1
{1,5}	2
{2,3}	3
{2,4}	2
{2,5}	2
{3,4}	-
{3,5}	1
{4,5}	-


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■ L2

Itemset	Support
{1,2}	3
{1,3} 	5
{1,5}	2
{2,3}	3
{2,4}	2
{2,5}	2

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■ Candidate C3

Itemset	Support
{1,2,3}	2
{1,2,5}	2
{1,2,4}	1
{1,3,5} 	1
{1,2,3,4}	-
{1,2,4,5}	1
{1,2,3,5}	-
{2,3,4}	-
{2,3,5}	1
{2,4,5}	-

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■ L3

Itemset	Support
{1,2,3}	2
{1,2,5}	2
{1,2,4}	1
{1,3,5}	1
{1,2,3,4}	-
{1,2,4,5}	1
{1,2,3,5}	-
{2,3,4}	-
{2,3,5}	1
{2,4,5}	-

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- **L3**

Itemset	Support
{1,2,3}	2
{1,2,5}	2

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- **Candidate C4**

Itemset	Support
{1,2,3,5}	1

- **To design association rule**

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- **Association rules**

Association Rules	Support	Confidence	Confidence %
$1 \rightarrow 2 \wedge 3$	2	$2/7 = 0.28$	28
$2 \rightarrow 1 \wedge 3$	2	$2/6 = 0.33$	33
$3 \rightarrow 1 \wedge 2$	2	$2/6 = 0.33$	33
$1 \rightarrow 2 \wedge 5$	2	$2/7 = 0.28$	28
$2 \rightarrow 1 \wedge 5$	2	$2/6 = 0.33$	33
$5 \rightarrow 1 \wedge 2$	2	$2/2 = 1.00$	100

The minimum confidence is 70%

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