

Bayesian Belief Networks in R

Abraham Mathew

Carmichael Lynch

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```
> head(dat, 10)
```

	Hair_Color	Race	Text	Pictures	Age	Miles_Away	Shared_Interest	Overweight	Liked_You
1	Black	White	Y	5	22 to 26	Close	0	N	N
2	Blonde	White	N	4	22 to 26	Close	1	N	N
3	Black	Other	Y	4	27 to 30	Close	4	N	N
4	Blonde	White	Y	5	22 to 26	Close	0	N	N
5	Blonde	White	N	4	18 to 21	Close	1	N	N
6	Brunette	White	Y	6	22 to 26	Close	0	N	N
7	Black	White	N	4	22 to 26	Close	1	N	N
8	Blonde	White	Y	6	22 to 26	Close	2	Y	N
9	Brunette	White	Y	6	18 to 21	Far	0	N	N
10	Blonde	White	Y	5	31+	Far	0	N	N

Review of Probability Theory

- Sample space: The set of all possible events of an experiment
- Marginal probability: the probability of an event occurring, or $P(A)$
- Conditional probability: the probability of event A occurring given that event B occurs, or $P(A|B)$
- Joint probability: the probability of event A and event B occurring, or $P(A \text{ and } B)$

Probabilistic Graphical Models

- Probabilistic graphical models represent the conditional dependencies between random variables through a graph structure.
- Nodes correspond to random variables and edges represent statistical dependencies between the variables.
- Bayesian Belief Networks (directed acyclic graphics) and Markov Network (undirected).

Bayesian Belief Networks

- Goal: learn the structure of a network and learn the parameters.



- This is a directed acyclic graphic (DAG) that shows the dependencies between the variables.
- The relationship between shared interests and liked you is dependent on the number of pictures.

Bayesian Belief Networks

- $P(LY|P)$, or what is the probability that a user liked you, given the number of pictures on their profile?
- Each node is associated with a probability function that specifies the probability of that variable conditional on the parent node.

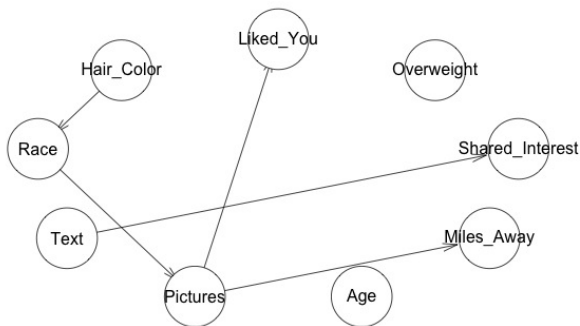
Pictures				
Liked_You	3	4	5	6
N	0.1666667	0.6363636	0.5384615	0.7500000
Y	0.8333333	0.3636364	0.4615385	0.2500000

BNLEARN Package

- bnlearn package (CRAN)
- The package implements a number of constraint based, pairwise, score-based, and hybrid structure learning algorithms.
- For the Tinder data set, use a score-based algorithm to learn the network structure.

```
library(bnlearn)  
net <- hc(dat)  
plot(net)
```


BNLEARN Package



BNLEARN Package

```
## HIGHLIGHT NODE AND ITS MARKOV BLANKET
plot(net, highlight = c("Pictures",
                        mb(net, "Pictures")))

## ESTIMATE PARAMETERS
fitted <- bn.fit(net, dat)
fitted

## STRENGTH OF THE ARCS
arc.strength(net, dat)

## CONDITIONAL PROBABILITY QUERIES
cpquery(fitted, (Liked_You=="Y"), TRUE)
cpquery(fitted, (Liked_You=="N"),
        (Pictures=="3" & Text=="N"))
```

Further Reading

- Ross, Sheldon. Introduction to Probability Models
- Scutari, Marco. Bayesian Networks: With Examples in R
- Nagarajan, Radhakrishnan, Scutari, Marco, and Lèbre, Sophie. Bayesian Networks in R: with Applications in Systems Biology