

HW1-R-and-pyro.ipynb

Causal inference with causal Bayesian networks bnlearn package and Pyro.ai

Note: This notebook contains both R and python code. Therefore, in order to run this notebook kernel and a python kernel configured in the jupyter notebook or jupyter lab session. The first part will run under an R kernel (with bnlearn and graphviz installed). The second part will run under a python

```
[1]: library(bnlearn)
```

Attaching package: 'bnlearn'

The following object is masked from 'package:stats':

sigma

Utility functions for displaying symbolic and numerical conditional distributions

```
[2]: display.cpd.chart <- function(dag, iss.list){
  for( node in nodes(bn.bayes)){
    for(iss in iss.list){
      bn.bayes <- bn.fit( dag, data= survey, method= "bayes", iss=iss)
      bn.fit.barchart(bn.bayes[[node]], main = paste(node.names[[node]],
        " (ISS=", iss, ")")
      xlab =display.cpd.str(dag, node) , ylab = "Probability mass")
    }
  }
}

display.cpd.str <- function( dag, node ){
  if( length(parents(dag, node)) == 0){
    paste("Pr(",node,")",sep="")
  } else {
    paste("Pr(",node,"|",paste(parents(dag,node), collapse=","),")", sep="")
  }
}
```

1 Building a DAG (4 points)

A DAG maps to a factorization of the joint distribution (e.g., $P(A, B, C) == P(A)P(B|A)P(C|A, B)$).