

# GMTK Tutorial

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# Outline

- GMTK Tutorial
  - What is GMTK.
  - Installation of GMTK
  - gmtkTriangulate and gmtkEMtrainNew
  - The GMTK Structure Language and master the file.
  - Example program
- Reference

# What is GMTK

- Graphical Models Toolkit
  - developed and maintained by University of Washington
- For developing dynamic Bayesian Network (DBN) based speech recognition and general time series systems.
- Open-source
  - But it hasn't released its source code yet.

# Why do we need to learn this tool

- Learning to use a tool is sometimes time-consuming.
  - Especially for those tools that are not well-documented.
  - But sometimes it's worth it.
- Each time we design a graphical model, we always want to know whether it can explain the data well.
  - But we have to implement it first.
- If we have tools on hand, we can spend much less time on building the model and know the performance of our model without lots of effort.

# Installation

- Download from
  - <http://ssli.ee.washington.edu/~bilmes/gmtk/linux/2009/>
- We use the Linux binaries. (Jan 2009)
  - If you want to work on the 217 workstations, you should download the binaries in m64 directory.
- For hw1, you need two binary files:
  - gmtkTriangulate and gmtkEMtrainNew

# gmtkTriangulate

- Create a triangulate file, which will be used for the main training program (gmtkEMtrainNew).
- For hw1, you only need to download it and set the right path in the provided Makefile.
  - or put it in “pgm\_hw1/bin/”

# gmtkEMtrainNew

- gmtkEMtrainNew is the main EM training program for GMTK.
  - EM is an algorithm which use iterative method to update and estimate parameters.
- gmtkEMtrainNew requires three input files
  - structure file (\*.str)
  - master file (\*.master)
  - data file (\*.pfile)

# GMTKL

- The GMTK structure language.
  - Specify the structure of a graphical model
  - Used in structure file. (\*.str)
- Example HMM structure file
  - You can find it in “pgm\_hw1/example/hmm.str”



# HMM: structure file (hmm.str)

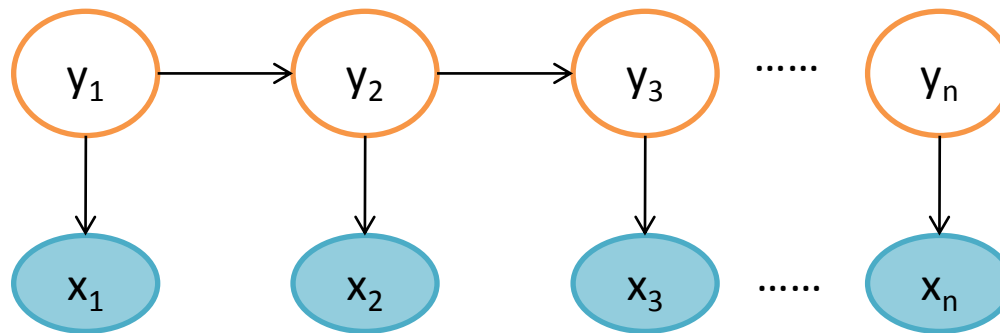
GRAPHICAL\_MODEL hmm

```
frame : 0 {  
  variable : state {  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: nil  
      using DenseCPT("InitialState");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0)  
      using DenseCPT("Emission");  
  }  
}
```

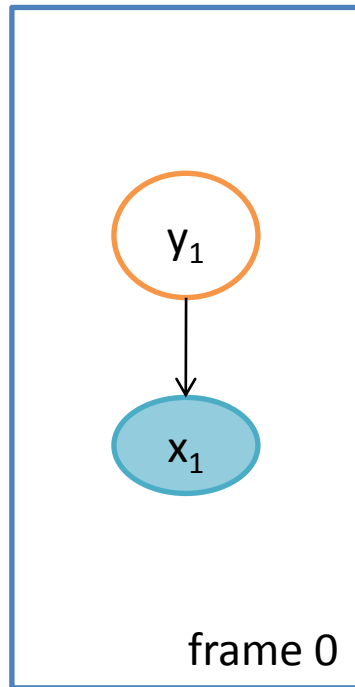
```
frame : 1 {  
  variable : state{  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(-1)  
      using DenseCPT("Transition");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0)  
      using DenseCPT("Emission");  
  }  
}  
chunk 1:1
```

# HMM: graphical structure

GRAPHICAL\_MODEL hmm

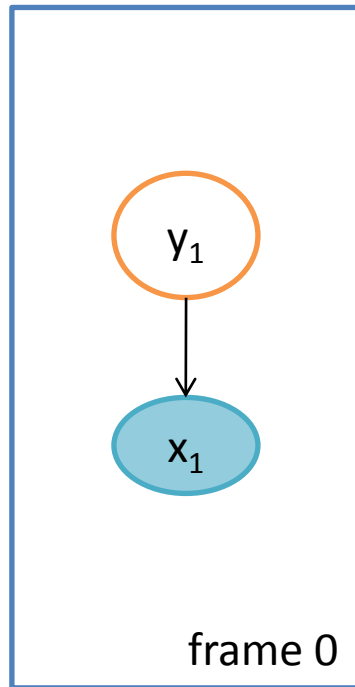


# HMM: graphical structure



```
frame : 0 {  
  y1 {  
    variable : state {  
      type: discrete hidden cardinality 2;  
      switchingparents: nil;  
      conditionalparents: nil using DenseCPT("InitialState");  
    }  
  }  
  x1 {  
    variable : outcome{  
      type: discrete observed 0:0 cardinality 2;  
      switchingparents: nil;  
      conditionalparents: state(0) using DenseCPT("Emission");  
    }  
  }  
}
```

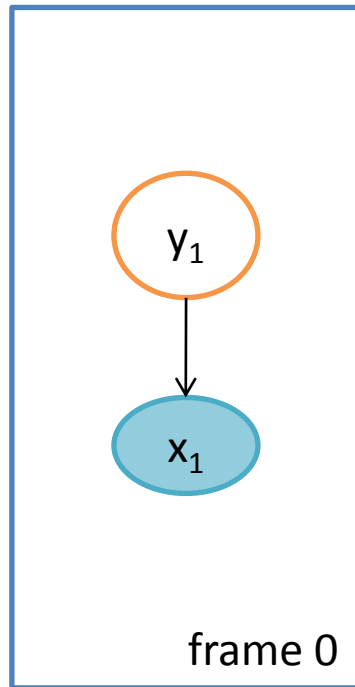
# HMM: graphical structure



```
frame : 0 {  
  variable : state {  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: nil using DenseCPT("InitialState");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0) using DenseCPT("Emission");  
  }  
}
```

$y_1 = \{0, 1\}$

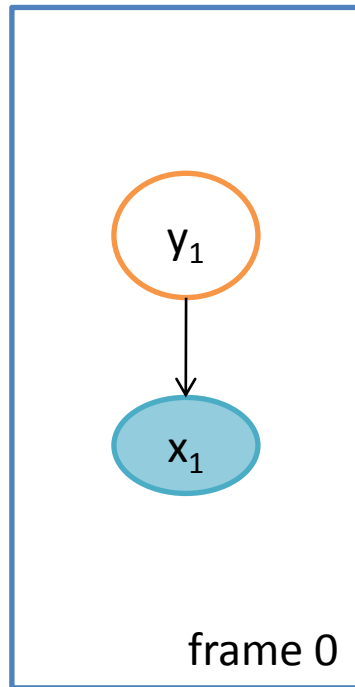
# HMM: graphical structure



```
frame : 0 {  
  variable : state {  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: nil using DenseCPT("InitialState");  
  }  
  variable : outcome {  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0) using DenseCPT("Emission");  
  }  
}
```

$y_1$  has no parent

# HMM: graphical structure



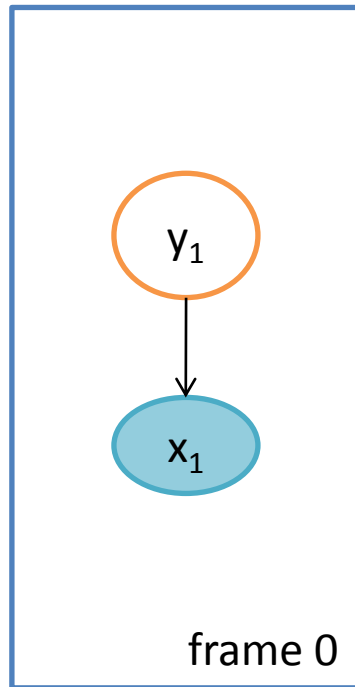
```
frame : 0 {  
  variable : state {  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: nil using DenseCPT("InitialState");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0) using DenseCPT("Emission");  
  }  
}
```

$y_1$  {

$x_1$  {

CPT of  $y_1$

# HMM: graphical structure



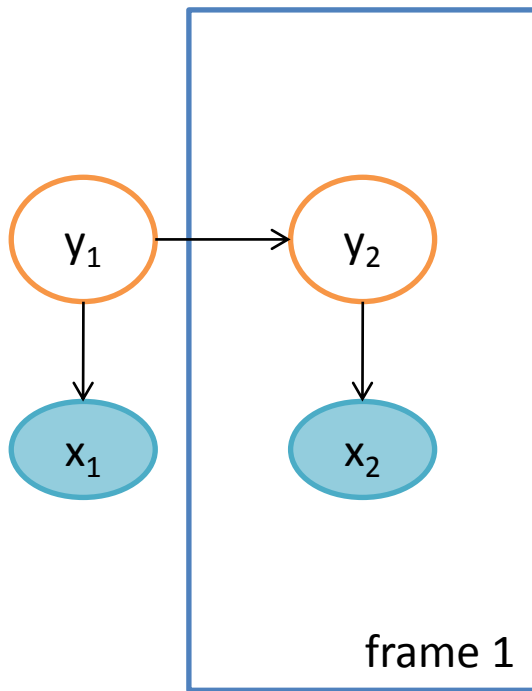
```
frame : 0 {  
  variable : state {  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: nil using DenseCPT("InitialState");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0) using DenseCPT("Emission");  
  }  
}
```

$y_1$

$x_1$

The dimension of this variable

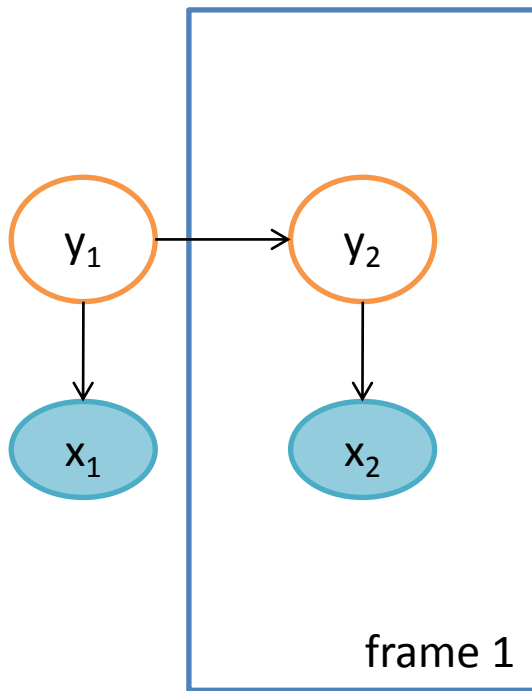
# HMM: graphical structure



```
frame : 1 {  
  variable : state{  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(-1) using DenseCPT("Transition");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0) using DenseCPT("Emission");  
  }  
}
```



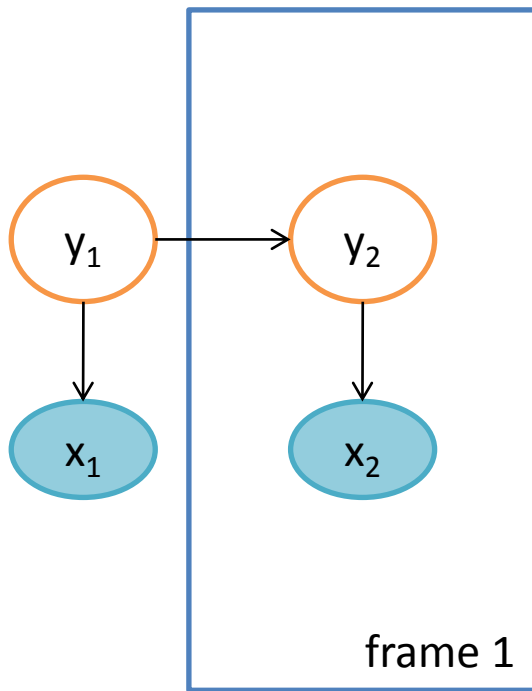
# HMM: graphical structure



```
frame : 1 {  
  variable : state{  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(-1) using DenseCPT("Transition");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0) using DenseCPT("Emission");  
  }  
}
```

Red annotations in the code block:  $y_2$  is bracketed to the `state` block,  $x_2$  is bracketed to the `outcome` block, and `state(-1)` is circled in red with a red  $y_1$  below it.

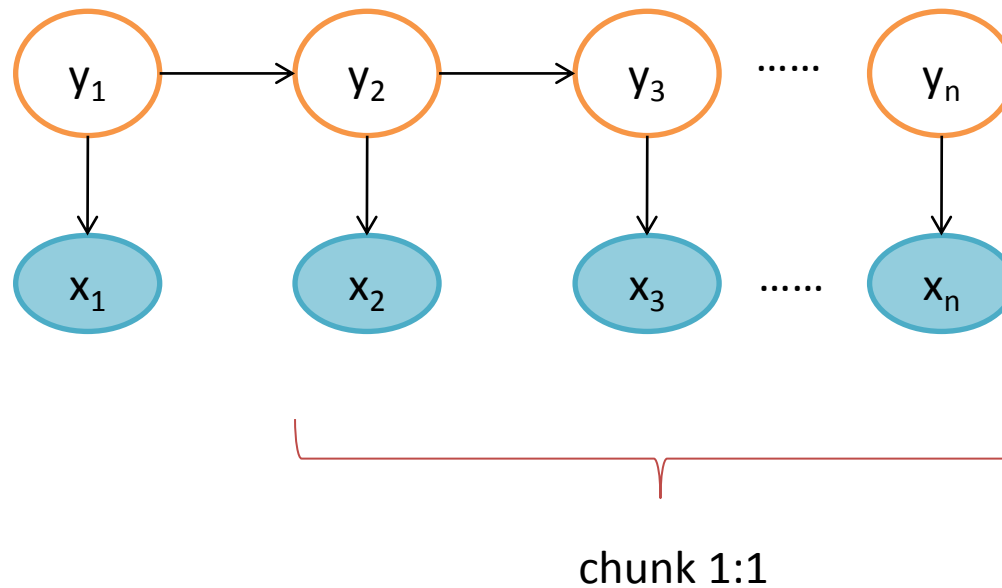
# HMM: graphical structure



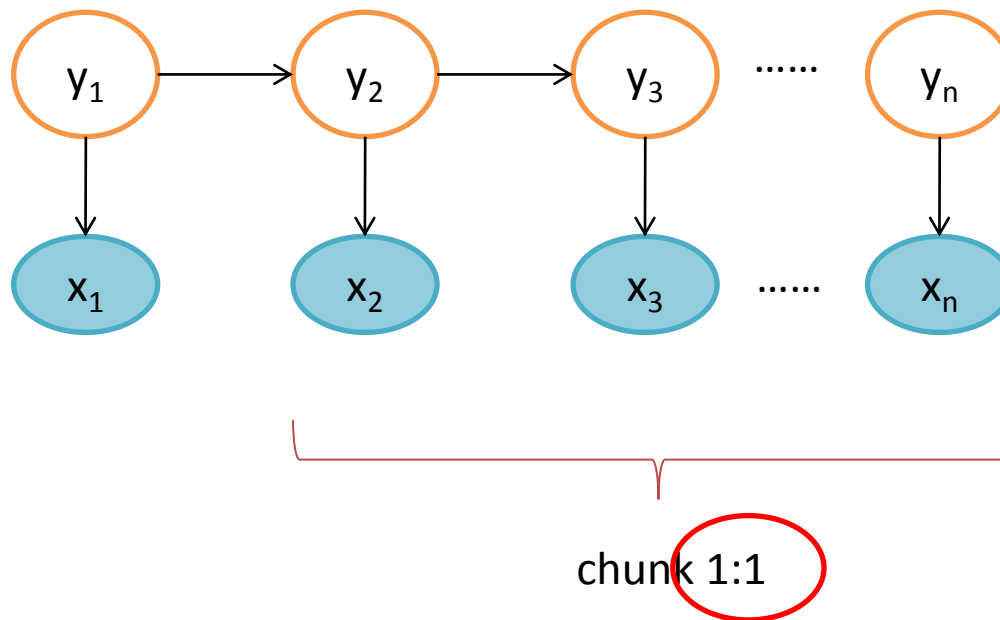
```
frame : 1 {  
  variable : state{  
    type: discrete hidden cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(-1) using DenseCPT("Transition");  
  }  
  variable : outcome{  
    type: discrete observed 0:0 cardinality 2;  
    switchingparents: nil;  
    conditionalparents: state(0) using DenseCPT("Emission");  
  }  
}
```

*CPT of  $y_2$*

# HMM: graphical structure



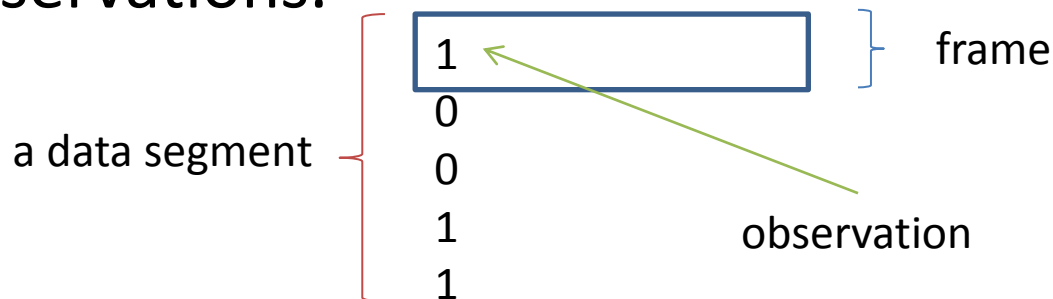
# HMM: graphical structure



means frame 1 to frame 1 is the repeating unit.

# HMM: data (hmm.pfile)

- This data file contains the observation of the graphical model.
  - There're two sections in pfile format
    - header and data
- There are multiple frames in a segment (data may have many segments), and each frame is a vector of observations.

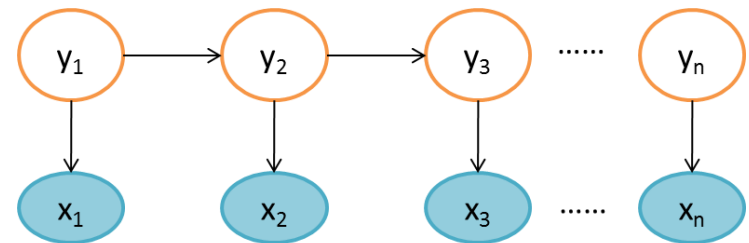


# HMM: master file

- The example master file (hmm.master) is empty.
  - because the CPTs are unknown, we want to learn them with EM.
- But what if we want to specify the CPTs?
  - Dense CPT

# Dense CPT: HMM master file

DENSE\_CPT\_IN\_FILE inline  
3 % number of dense CPTs  
0 % the first one  
InitialState % name of CPT  
0 % number parents  
2 % self cardinality = 2  
2.25058877030877800e-59 1.0000000000000000e+00  
1 % the second one  
Emission % name of CPT  
1 % number parents  
2 2 % parent cardinality = 2, self cardinality = 2;  
9.99999940233369022e-01 5.97666310031556278e-08  
3.48122867653281989e-07 9.99999651877132356e-01  
2 % the third one  
Transition % name of CPT  
1 % number parents  
2 2 % parent cardinality = 2, self cardinality = 2;  
9.09090862189374960e-01 9.09091378106250125e-02  
1.33333318275715290e-01 8.66666681724284738e-01



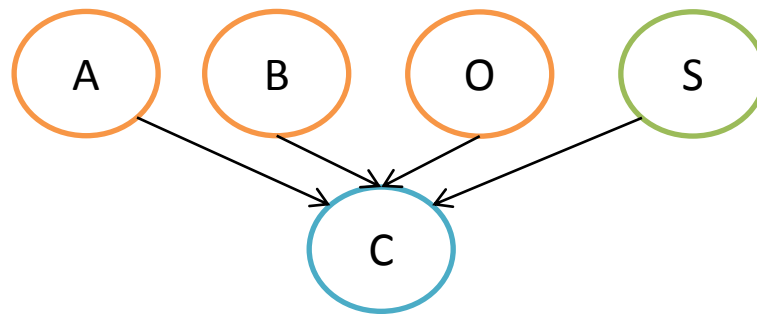
# Master file

- If you want to fix some parameters when you run `gmtkEMtrainNew`,
  - add `[-objsNotToTrain filename]` argument
  - otherwise they will be updated during the training.
- For example, if you add one line “DENSECPT InitialState” to the not-to-train file, the CPT of InitialSate will be fixed during the training.



# Generalized CPT using Decision Tree

- Assume there's a Bayesian Network with five nodes



- And we want to express an OR relationship  $\begin{cases} C=A, \text{ if } S=0 \\ C=B, \text{ if } S=1 \\ C=0, \text{ others} \end{cases}$
- We can use generalized CPT to represent it.
  - That is, we use a decision tree to define a CPT.
  - It may be helpful for hw1.

# str File for the BN

GRAPHICAL\_MODEL OrModel

```
frame : 0{
```

```
...
```

```
variable : C{
```

```
  type: discrete observed 0:0 cardinality 2;
```

```
  switchingparents: nil;
```

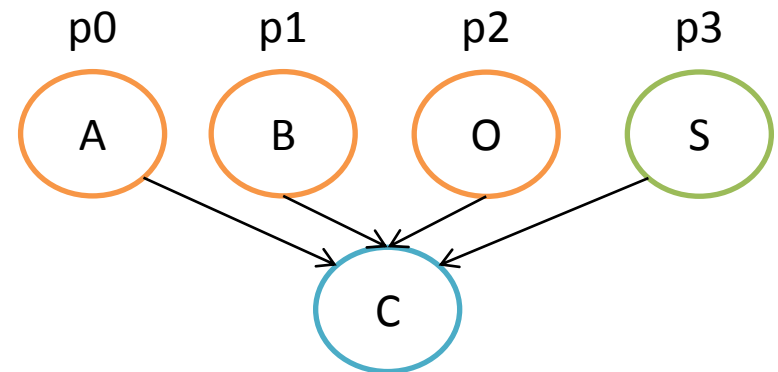
```
  conditionalparents: A(0), B(0), O(0), S(0)
```

```
    using DeterministicCPT("orCPT");
```

```
}
```

```
}
```

```
chunk 0:0
```



# master file for the BN

DETERMINISTIC\_CPT\_IN\_FILE inline

1 % number of deterministic CPT

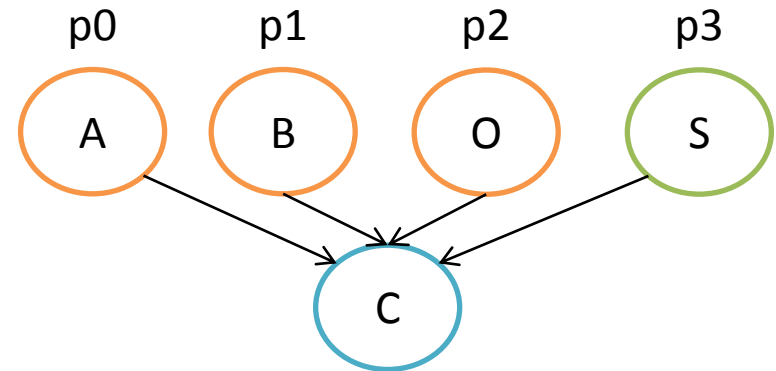
0 % the first one

orCPT % name of CPT

4 % number of parent

3 3 3 2 3 % cardinality of A, cardinality of B, cardinality of O, cardinality of S, cardinality of C

orDT % name of DT used for this CPT



DT\_IN\_FILE inline

1 % number of decision tree

0 % the first one

orDT % name of DT

4 % number of parent

3 3 0 1 default % (query parent, number of splits, integer range, integer range, "default")

-1 {{p0}} % first split, p0: the value of the first parent

-1 {{p1}} % second split, p1: the value of the second parent

-1 {{p2}} % default split, p2: the value of the third parent

# Example program

- The example program is in “pgm\_hw1/example”
  - Type the command “make train” to start training.
  - The learned CPTs can be found in hmm.cpt.
  - The structure file is hmm.str.
- You can see the Makefile for the detailed arguments.

# Reference

- GMTK documentation files
  - <http://ssli.ee.washington.edu/~bilmes/gmtk/doc.pdf>

# PGM Hw1 part1

- Two hand-written questions and two programming problems.
- Deadline:
  - Hand-written section: 10/9 (Tue) before class
  - Programming section: 10/14 (Sun) 23:59