

$$P(x_1, x_2, \dots, x_d) = \prod_{j=1}^d P(x_j | x_{1:j-1})$$

if $\forall i, n_i \in \{0, 1\}$ \Rightarrow need 2^d parameters

Simplification

$$P(x_j | x_{1:j-1}) = P(x_j | \underbrace{x_{\text{pa}(j)}}_{\text{parents of } j})$$

$$\underline{x_j \perp x_{-\text{pa}(j)} | x_{\text{pa}(j)}} \rightarrow$$

$\Rightarrow 2^{(\# \text{pa}(j)+1)}$ parameters

$\#(\text{parents of } j)$

x_j is independent of the other n_i given its parents

DAG \rightarrow to represent conditional

$G = (V, E)$ dependencies

$V \rightarrow n_i ; E \rightarrow \text{if } (v, v) \in E$

$\Rightarrow v$ is a parent of v

D-separation rules

1) $(\textcircled{x}) \quad (\textcircled{y}) \Rightarrow x \perp y$

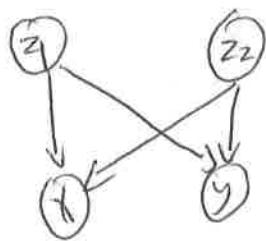
2) $(\textcircled{x}) \rightarrow (\textcircled{y}) \Rightarrow x \not\perp y$

3)

 $\Rightarrow x \not\perp y$

4)

 $\Rightarrow x \not\perp y | z$



$$n \perp y \mid z_1, z_2$$

$$n \not\perp y \mid z_1$$

$$n \not\perp y \mid z_2$$

5)

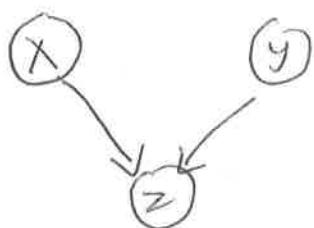
6)



$$n \not\perp y$$

$$n \perp y \mid z$$

7)

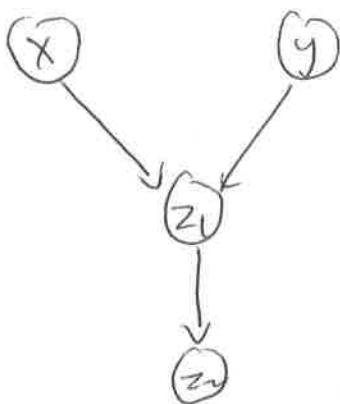


$$n \perp y$$

$$n \not\perp y \mid z$$

V-structure

8)

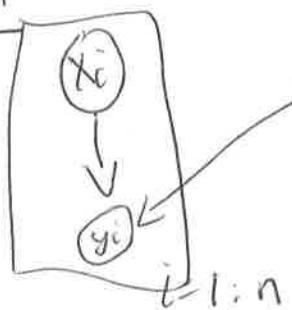


$$n \perp y$$

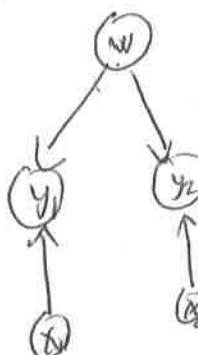
$$n \not\perp y \mid z_2$$

Plate Notation

MLE



\Rightarrow



$$y^i \sim N(w^T x^i \mid, 1)$$

Project ideas

1) Kaggle / crowdAnalytiX

- US elections 2016 / NcAA basketball
- SF crime / Yelp photos → Multilabel classification / Home depot → Sand relevance
- multi-class

2) Survey

- stochastic gradient methods
- transfer learning
- Bayesian optimization
- Multitask learning / multi-label
- Tensor factorization / spectral methods
- Sketching techniques
- classification / anomaly detection for streaming data

3) Coding project

- Spectral clustering / ICA / Probabilistic PCA / (code up some theory paper & check how practical it is)

4) Theory

- matrix completion
- active learning
- 2 layer NN's

- submodular maximization
- bandits
- online learning
- randomized SVD

5) check recent AISTATS/NIPS/ICML papers