

Graphical Models

GM/FAT

11/05/2010

x_1, x_2, \dots, x_n

$t=0$ $x_1^0, x_2^0, \dots, x_n^0$

$t=1$ $x_1^1, x_2^1, \dots, x_n^1$

\vdots

\vdots

$\rightarrow P(x_1, \dots, x_n) = ?$

2 values

$x_1^0, x_2^0, \dots, x_n^0$	$P = \frac{1}{10}$
$x_1^1, x_2^1, \dots, x_n^1$	$P = \frac{1}{100}$
\vdots	\vdots

$\rightarrow 2^n$

\downarrow

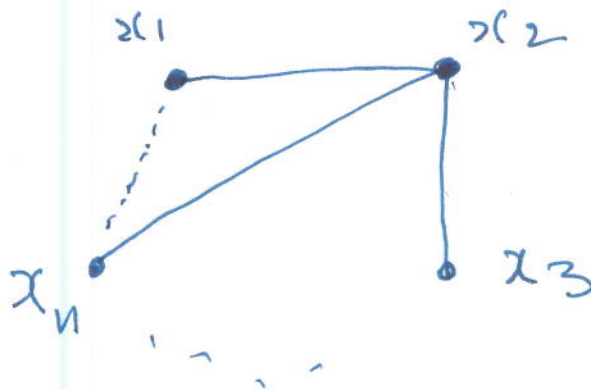
\downarrow

of samples

$\gg 2^n$

2^{2^n}
 $\downarrow O(n^4)$

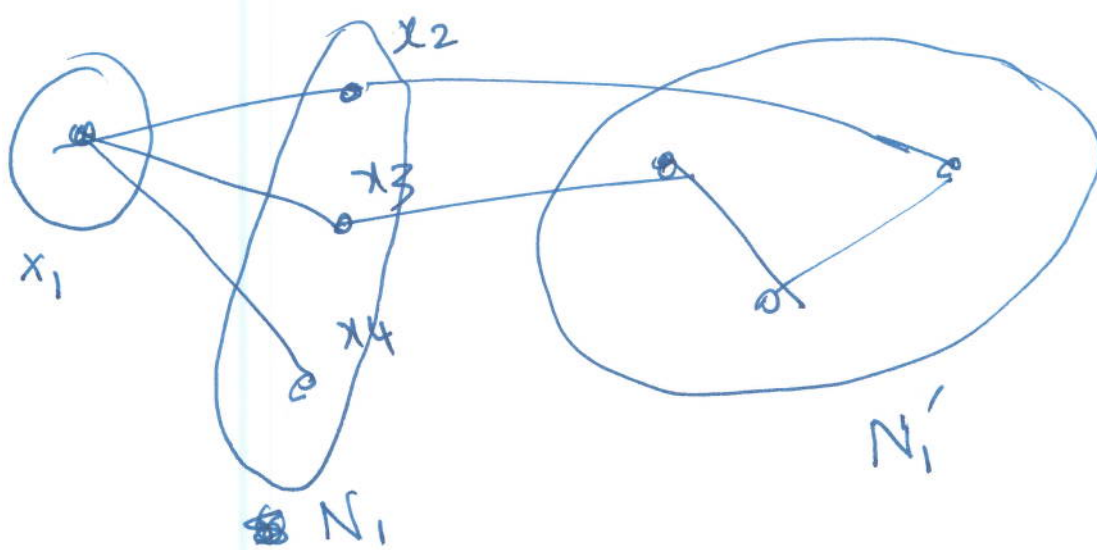
Suppose, I know the distribution



$$P[x_1 | x_2, \dots, x_n]$$

=

$$P[x_1 | x_2]$$



$$P(x_1 | N_1, N_1') = P(x_1 | N_1)$$

2^n	(x_1, x_2, \dots, x_n)	P_i

2^d	(x_1, N_1)	P_i

$$\left\{ \begin{array}{l} (x_1^{(i)}, x_2^{(i)}, \dots, x_p^{(i)}) \quad \text{for } i=1, \dots, n \\ \text{find } P(x_1, \dots, x_p) = ? \end{array} \right.$$

It suffices to find the graphical Model.

parametric fns : $P = f(\theta_1, \dots, \theta_k)$

$$\frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$f(\sigma, \mu)$$

non-parametric fns : $P = \cancel{f(\theta_1, \dots, \theta_k)}$

Assumption: P is parametric

~~P~~

$$P[x_r = j \mid X_{1:r} = x_{1:r}] = \frac{\exp(\theta_{r,j} + \sum_{l=1}^{r-1} \theta_{r,l;j} I(x_l = k))}{1 + \sum_{k=1}^m \exp(\theta_{r,j} + \sum_{l=1}^{r-1} \theta_{r,l;j} I(x_l = k))}$$

$\exp(\theta_{r,j} + \sum_{l=1}^{r-1} \theta_{r,l;j} I(x_l = k))$

$$P[X_r=j | X_t=k; X \dots] \propto \exp(\theta_{r;j} + \theta_{r;t;jk})$$

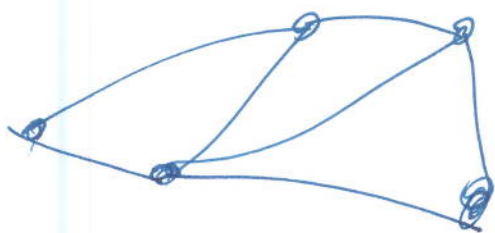
$\underbrace{\theta_{r;j}}_{\text{nodes}}$ $\underbrace{\theta_{r;t;jk}}_{\text{values}}$

$$\theta_{r;t;jk} \neq 0 \Rightarrow$$


suppose for all k $\theta_{r;t;jk}$'s are equal



Theory Sparsity: Markov Random Fields

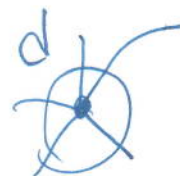


of edges $\ll p^2$

$O(\log p)$ \downarrow $O(p)$

$$\frac{dp}{2}$$

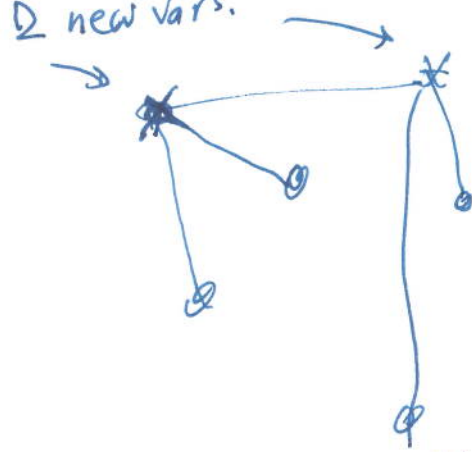
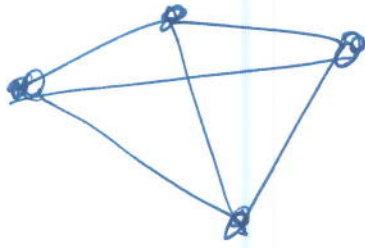
bounded degree



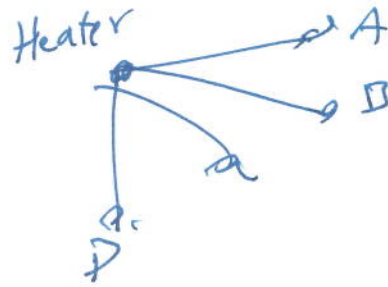
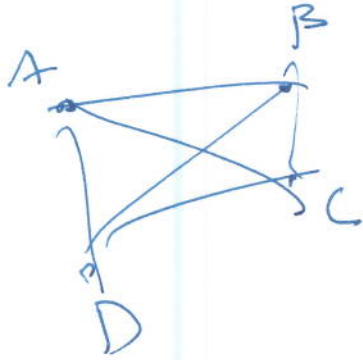
p nodes

$O(p^2)$ edges

$$\binom{p}{2}$$



Latent variables



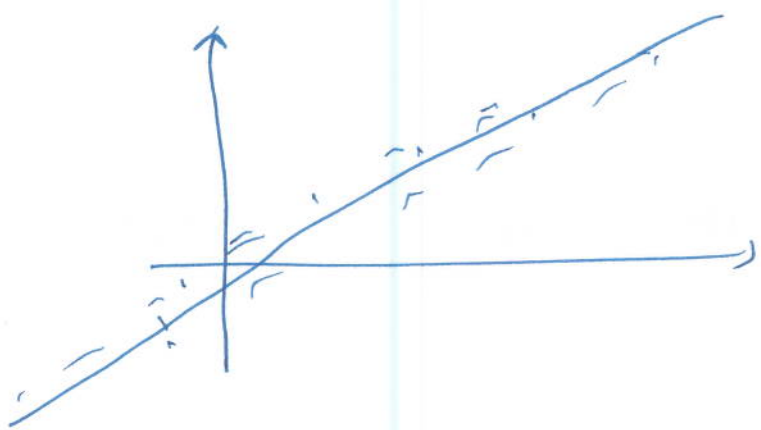
→ $\theta_{r;ij}$'s

$\theta_{rt;jk}$'s



I WANT: most of them to be zero

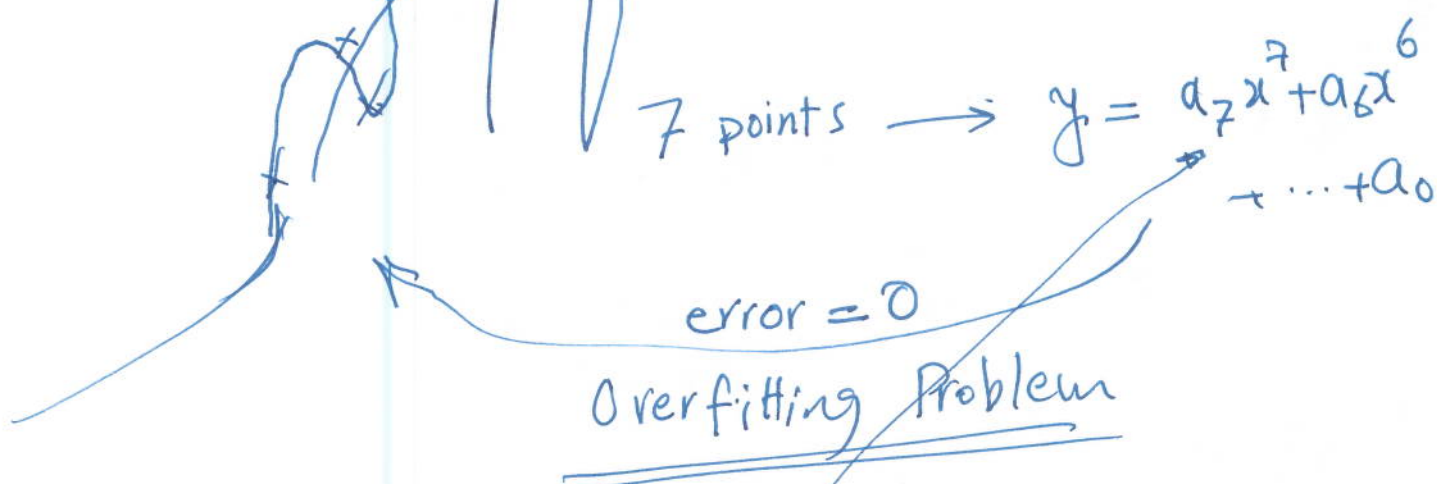
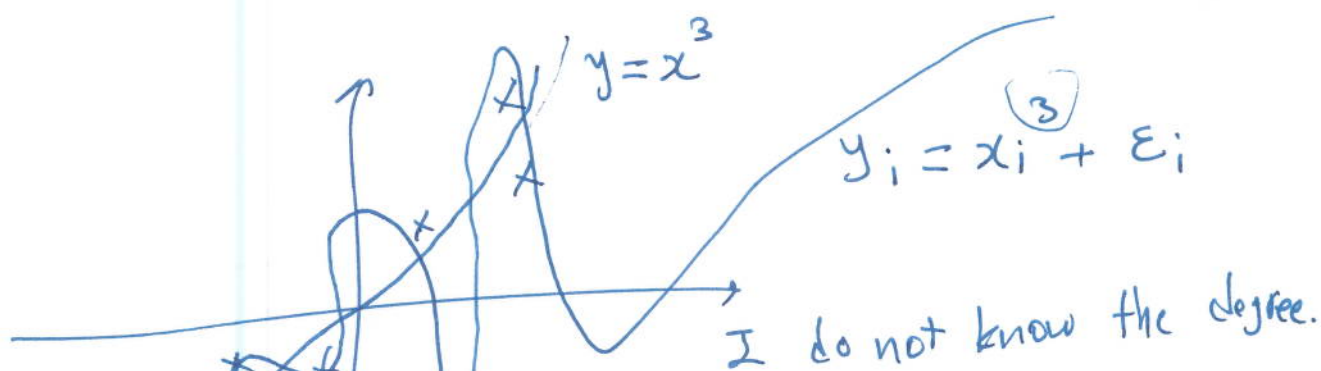
LASSO → Regression



$$y = ax$$

$$\begin{matrix} y_1 & x_1 \\ y_2 & x_2 \\ \vdots & \vdots \end{matrix}$$

$$\min_{a \in \mathbb{R}} \frac{1}{n} \sum_{i=1}^n \|y_i - ax_i\|_2^2 \quad \leftarrow \quad y_i = ax_i + \varepsilon_i$$



$$\min_a \frac{1}{n} \sum_{i=1}^n \|y_i - a_7 x_i^7 - a_6 x_i^6 - \dots\|_2^2 + \lambda \|a\|_1 \rightarrow \begin{matrix} a_7 = 0 \\ a_6 = 0 \\ a_5 = 0 \\ \vdots \end{matrix}$$

gain \leftarrow ~~price~~

price \leftarrow $\sum |a_i|$

λ_3

λ_3 \leftarrow λ_3 \leftarrow $\lambda_3 = \lambda_1 \oplus \lambda_2$