## Dynamic Programming & Graphical Model 2/8/19

## Graphical Models for prob. dist.

A way to define high dimensional prob dist.

Factor Graphs on example first

Ex. Medical diagnois

C={ ) if I have a cold ow.

S= { ) if I have a sure throat

Y= { ) ; f I have a runny nose 0 o.w

P= { 1 if there is pollen in the crir

	C	5	~	Y, (c,s,r)	<u> </u>	,	P	46(r,p)
•				.1	_	_		١
				,3		] ]	D 1	.2
	_			,6		$\circ$	Ö	.4

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Assume Yal 45 ind unless told O.W.

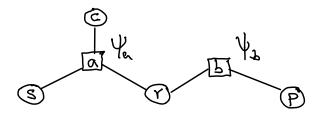
in  $P_{rob}(c=0, s=0, v=1, P=1) = (3)(.3) = .09$  $P(c,s,r,p) = \frac{1}{7} \gamma_q(c,s,r) \gamma_b(r,p)$ 

2 = normalization constant.

In our example

Such be 25 states
The remaining ones are U prob.

View as a graph = factor graph



In general i Factor Graph is a hipportite

The graph will have:

- 1) variable nodes: V's
- 2) factor nodes 3 es da = neig of a where da = variables of Ya.
- 3) for each factor a map:

  Y: {v,1}Bal ~ R+

Tasks: Giren en factor graph G.

- 1) Computer marginals in P(X;=1)
- 2) Compute conditionals in P(X2=1 ] Xi=0)
- 3) Same from dist.
- 4) Find mode ie argmax {P(x)}

Claim For general factor graph computing any of the above in NP-hard.

We will show: 3-SAT ≤p Marginal the others are similar.

<u>Aecall</u>: 3-(NF is a Boolean formula with 3 lituals per clause.

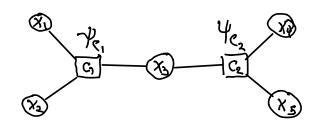
3-SAT = { pe3-(NF | p has a satisfing instance)

es  $F=(\chi_1 \vee \chi_2 \vee \chi_3) \wedge (\chi_3 \vee \chi_4 \vee \chi_5) \in 3\text{-CNF}$ Note  $(\chi_1 \vee \chi_2 \vee \chi_3) \text{ is false iff } \chi_1 = \chi_2 = f \text{ if } \chi_3 = f \text{ if } \chi_4 = \chi_5 = f \text{ if } \chi_5 = f \text{ if$ 

Suppose DE3-CNF with clauses Ci, ..., Cm. for each clause c Let 4c be the factor s.t. 4 = & 1 if true (0 o.w.

In our example:

 $Y_{C_1}(X_1, X_2, X_3) = \begin{cases} 1 & \text{if true} \\ 0 & 0, w \end{cases}$ 



Consider uniform dist over {0,1}5

 $P(x_3 - x_5) = \frac{1}{2} \Upsilon_{C_1}(x_1, x_2, x_3) \Psi_{C_2}(x_3, x_4, x_5)$ 

When Z ≠ O Pio uniform dist over gatisting assignments

Thus p(X1=1)>U iff Fis satisfiable.
Thus computing marginals is NP-Hard

Note Computing Z is #P-Complete.

We restrict factor graphs to a tree.

To show For trees factor graph tasks

1),--,4) are poly-time.

We will solve the unnormalized prob

 $Z[\chi_{i}=b] = \sum_{\substack{\chi \in \{0,1\}^{n} \\ \chi_{i}=b}} \prod_{\alpha=i}^{m} \Psi_{\alpha}(\chi_{\partial\alpha})$ 

Note: P[X; = b] = Z[X; = b]/Z

Goal: Compute Z[X;=b] We root G at vertex X; Subproblems: Rooted subtrees of G. Lt Tw be subtree vooted at w (var infait) two types of subtrees. 1) Variable rooted a) Factor vooted

Tx be subtree of Tx with child ex Vn be all varieble in Tw (Tw)

Fn he all factors in Tw (Tw)

Child(w) = children (w)

For Variable-vooted .

Yieln1 & Y be {0,1}

compute: V; (b) = \( \sum\_{\chi\_1} \)

compute. Vi(b) = \( \sum\_{\ext{\end}} \forall \forall

## For factor-rooted trees:

∀ienn & ∀S ∈ ευ,13 & ∀Œ ∈ child(xi) Computa:

$$V_{i}(\bar{a},b) = \sum_{\kappa} Y_{\kappa}(Y_{3\bar{a}}) \prod_{\kappa} Y_{\kappa}(Y_{3\bar{a}}) \prod_{\kappa$$

## 3) Recurrences:

For variable-rooted subtrees

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For factor-rooted subtrees

$$V_{i}(\hat{a}_{j}b) = \begin{cases} \sum_{\gamma \in \{0,1\}} \partial \bar{a}_{i} & \forall_{i} \forall_{j} \forall_{j} \forall_{i} \forall_{j} \\ \forall_{i} \forall_{i} \forall_{j} \end{cases}$$

$$V_{i}(\hat{a}_{j}b) = \begin{cases} \sum_{\gamma \in \{0,1\}} \partial \bar{a}_{i} & \forall_{i} \forall_{j} \forall_{j} \forall_{j} \forall_{j} \forall_{j} \forall_{j} \forall_{i} \forall_{$$

4) Correction: Simple induction

Called the Sum-Product Alg