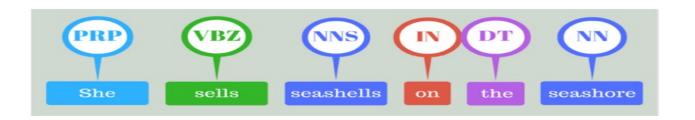
Sequence Labeling & Conditional Random Field (CRF)

References:

- https://cocoxu.github.io/CS7650 fall2021/slides/le c6-seq2.pdf
- https://www.youtube.com/watch?v=2KTeXhfsc-k

Sequence Labeling Use Cases



When Sebastian Thrun PERSON started working on self - driving cars at Google ORG in 2007 DATE , few people outside of the company took him seriously . " I can tell you very senior CEOs of major American NORP car companies would shake my hand and turn away because I was n't worth talking to , " said Thrun PERSON , in an interview with Recode ORG earlier this week DATED .

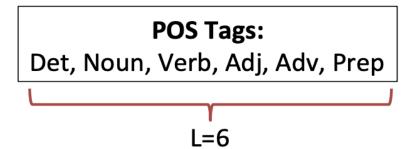
Q: Who is the first wife of Albert Einstein?

E: Einstein/O married/O his/O first/O wife/O eva/B Marić/I in/O 1903/O

A: Mileva Marić

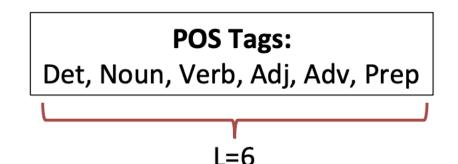
Sequence Labeling: POS

- Input: $x = (x^1, ..., x^M)$
- Predict: $y = (y^1, ..., y^M)$
 - Each yⁱ one of L labels.
- x = "Fish Sleep"
- y = (N, V)
- x = "The Dog Ate My Homework"
- y = (D, N, V, D, N)
- x = "The Fox Jumped Over The Fence"
- y = (D, N, V, P, D, N)



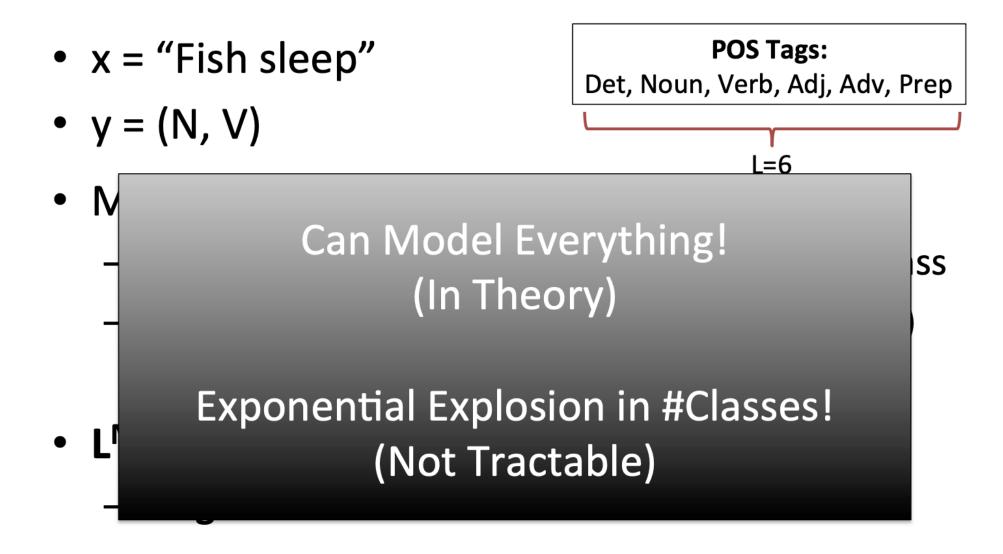
General Multi-class

- x = "Fish sleep"
- y = (N, V)



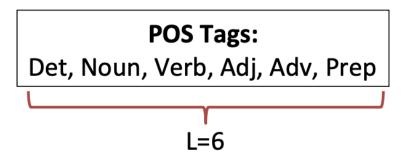
- Multiclass prediction:
 - All possible length-M sequences as different class
 - (D, D), (D, N), (D, V), (D, Adj), (D, Adv), (D, Pr) (N, D), (N, N), (N, V), (N, Adj), (N, Adv), ...
- L^M classes!
 - Length 2: $6^2 = 36!$

General Multi-class



Independent Multiclass

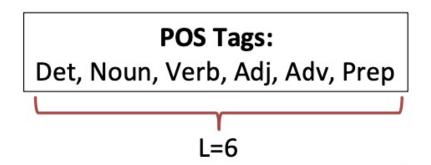
x="I fish often"



- Treat each word independently (assumption)
 - Independent multiclass prediction per word
 - Predict for x="I" independently
 - Predict for x="fish" independently
 - Predict for x="often" independently
 - Concatenate predictions.

Independent Multiclass

x="I fish often"

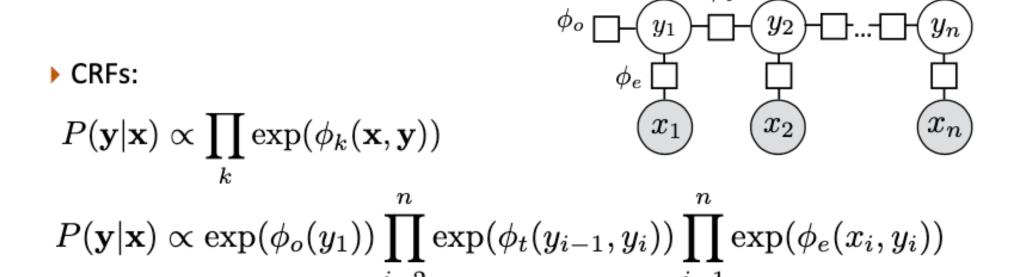


#Classes = #POS Tags (6 in our example)

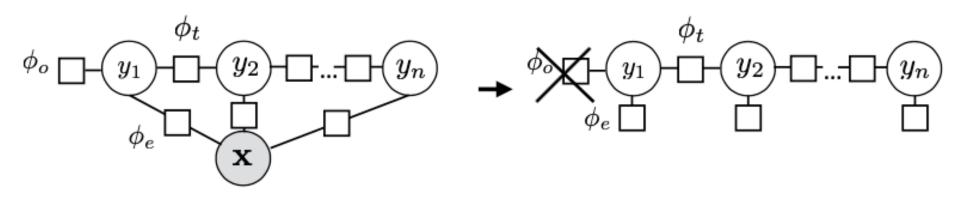
Solvable using standard multiclass prediction.

But ignores context!

Sequential Conditional Random Fields (CRF)



Sequential CRFs



- Notation: omit x from the factor graph entirely (implicit)
- Don't include initial distribution, can bake into other factors

$$P(\mathbf{y}|\mathbf{x}) = \frac{1}{Z} \prod_{i=2}^{n} \exp(\phi_t(y_{i-1}, y_i)) \prod_{i=1}^{n} \exp(\phi_e(y_i, i, \mathbf{x}))$$

Feature Functions

$$P(\mathbf{y}|\mathbf{x}) = \frac{1}{Z} \prod_{i=2}^{n} \exp(\phi_t(y_{i-1}, y_i)) \prod_{i=1}^{n} \exp(\phi_e(y_i, i, \mathbf{x})) \underbrace{\begin{pmatrix} y_1 \\ y_2 \end{pmatrix}}_{\phi_e} \underbrace{\qquad \qquad \qquad }_{\phi_e} \underbrace{\qquad \qquad \qquad }_{\phi_e} \underbrace{\qquad \qquad }_{\phi$$

Phis can be almost anything! Here we use linear functions of sparse features

$$\phi_e(y_i, i, \mathbf{x}) = w^{\top} f_e(y_i, i, \mathbf{x}) \quad \phi_t(y_{i-1}, y_i) = w^{\top} f_t(y_{i-1}, y_i)$$
$$P(\mathbf{y}|\mathbf{x}) \propto \exp w^{\top} \left[\sum_{i=2}^n f_t(y_{i-1}, y_i) + \sum_{i=1}^n f_e(y_i, i, \mathbf{x}) \right]$$

Looks like our single weight vector multiclass logistic regression model

Features for NER

- Word features (can use in HMM)
 - Capitalization
 - Word shape
 - Prefixes/suffixes
 - Lexical indicators
- Context features (can't use in HMM!)
 - Words before/after
 - Tags before/after
- Word clusters
- Gazetteers

Leicestershire

Boston

Apple released a new version...

According to the New York Times ...