

## Quiz 4

	e NER task, when the the suffix feature "-field" fires, it is actually quite ve. All else being equal, what is a word that has the suffix "field" most likely
	Person name
	Organization
_	Movie name
	Location
	Not a named entity
	t is NOT true about MaxEnt models?
	Also known as log-linear models
	It is trying to maximize the entropy while respecting observed evidences
	The weights for features need not sum up to 1 but must lie in $[0,1]$ .
	None of the above
_	h of the following is true?
	An HMM is better than a MEMM because we can easily add arbitrary features
_	n HMM.
	The disadvantage of a discriminative model is that training and optimizing parameter weights is more expensive than for a generative model.
<b>[</b>	A trigram language model is making a Markov assumption that the probability of a word depends on all the words that appear before it.
_	All of the above
	None of the above
4: A Ma normali	eximum Entropy model takes the exponential "vote" for each class and the exponential to get a probability (see lecture notes). If there is a binary ation problem, and $vote(c_1) = 0.2$ and $vote(c_2) = -0.2$ , which of the following
	$p(c_1) + p(c_2) = 1$
	$ p(c_1) > p(c_2) $
_	Both of the above
	None of the above

5: Which of the following is true?
Character substrings and word shapes are very indicative features for the NER
task.
A CMM (a.k.a. MEMM) makes a single decision at a time, conditioned on
evidence from observations and also previous decisions.
If we apply a MaxEnt classifier (a non-sequence model) to an NER task, then features can be from the observed context but not from what labels (NER tags)
they have.
All of the above
None of the above
6: In a MaxEnt model, a feature C is formed as the intersection of two features A
and B. If we are trying to predict many classes and using many features at once, then, when looking at the specific weights that A, B, and C have for a specific class,
which of the following statements is true:
If A and B agree in sign, C will have the same sign.
C will have the sign of whichever A or B was greater in magnitude.
If A and B agree in sign, C will have the opposite sign.
C could have either sign in general.  7: For an HMM parameterized by A and given a set of observations, O, the problem
of finding $P(A O)$ is known as:
Likelihood
Decoding
Learning
Encoding
None of the above
8: Which of the following graphs describe a generative model:
On the second of the second of grant and grant a
This sentence might be labeled

labeled

This

sentence

might

be

9: Which of the following does not accomplish proper regularization:

$$\arg\max_{w} \sum \log p(y|x) - ||w||^{2}$$

$$\arg\max_{w} \prod p(y|x) \exp\left(\frac{w^{T}w}{2\sigma^{2}}\right)$$

$$\arg\max_{w} \sum \log p(y|x) - \sum |w|$$

$$\arg\max_{w} \prod \frac{p(y|x)}{\max\{1, |w|\}}$$

$$\Box a$$

$$\Box b$$

$$\Box c$$

$$\Box d$$

All of them accomplish proper regularization

10: In MaxEnt models, we are trying to estimate the parameters w as:

$$\arg\max_{w}\prod_{(c,x)\in(C,X)}p(c|x,w)$$

which is equivalent to:

$$\arg\max_{w} \sum_{(c,x) \in (C,X)} \log p(c|x,w)$$

Recall that the final form of p(c|x,w) in a MaxEnt model is:

$$p(c|x,w) = \frac{\sum_{i} w_{ci} f_i(c,x)}{\sum_{c'} \exp\left(\sum_{i} w_{c'i} f_i(c',x)\right)}$$

Finding the optimal weights that satisfy the above equations then guarantees that the derivatives of this function are zero for all weights  $w_{c-hat,j}$ , or:

$$\frac{\partial}{\partial w_{\hat{c}j}} \sum_{(c,x) \in (C,X)} p(c|x,w) = 0$$

This condition then guarantees that which terms are equal:

$$\sum_{(c,x)\in(C,X)} f_j(c,x) = \sum_{(c,x)\in(C,X)} f_j(\hat{c},x) p(\hat{c}|x,w)$$

$$\sum_{\{(c,x)\in(C,X): c=\hat{c}\}} f_j(c,x) = \sum_{(c,x)\in(C,X)} f_j(\hat{c},x) p(\hat{c}|x,w)$$

$$\sum_{(c,x)\in(C,X)} f_j(c,x) = \sum_{\{(c,x)\in(C,X): c=\hat{c}\}} f_j(\hat{c},x) p(\hat{c}|x,w)$$

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None of the above