

עיבוד שפה טבעית - תרגיל בית 2

מגשים:
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שאלה 1: בדפים הסרוקים

שאלה 2:
סעיפים a,b: בדפים הסרוקים

סעיף d:

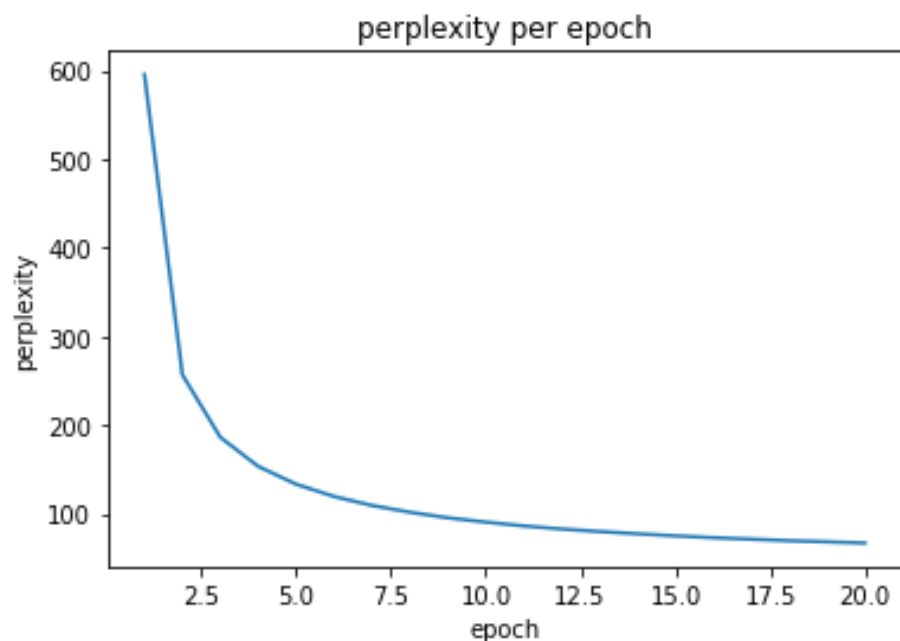
תוצאת הריצה בשלב ה eval:

Perplexity = 112.967665327491

שאלה 3:

סעיף a: בדפים הסרוקים

סעיף c:
גרף ערכי ה perplexity לאורך האימון:



ערך ה perplexity הסופי: 104.709

שאלה 1 b:

$\lambda_1 \backslash \lambda_2$	0	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1
0	207	424	401	87	78	77	68	64	62	61	
0.1	111	87	76	69	64	60	57	56	55		
0.2	92	75	67	62	58	55	54	53			
0.3	83	68	61	57	54	53	52				
0.4	77	64	58	55	53	52					
0.5	75	61	56	54	53						
0.6	74	60	56	54							
0.7	75	60	57								
0.8	80	63									
0.9	93										
1											

Best Parameters: $\lambda_1 = 0.4$

$\lambda_2 = 0.5$

Perplexity = 52.43

$$CE(y, \hat{y}) = - \sum_i y_i \log(\hat{y}_i)$$

(a) (2)

$$\text{Softmax}(\theta)_i = \frac{\exp(\theta_i)}{\sum_j \exp(\theta_j)}$$

$$\frac{\partial CE(y, \hat{y})}{\partial \theta_i} = \frac{\partial}{\partial \theta_i} \left(- \sum_k y_k \log(\hat{y}_k) \right) =$$

$$= \frac{\partial (- \sum_k y_k \log(\hat{y}_k))}{\partial \log(\hat{y}_k)} \cdot \frac{\partial \log(\hat{y}_k)}{\partial \hat{y}_k} \cdot \frac{\partial \hat{y}_k}{\partial \theta_i} =$$

$$= - \sum_k \frac{\partial (y_k \log(\hat{y}_k))}{\partial \log(\hat{y}_k)} \cdot \frac{1}{\hat{y}_k} \cdot \frac{\partial \hat{y}_k}{\partial \theta_i} =$$

$$= - \sum_k y_k \cdot \frac{1}{\hat{y}_k} \cdot \frac{\partial \hat{y}_k}{\partial \theta_i} = \dots$$

$$\frac{\partial \hat{y}_k}{\partial \theta_i} = \frac{\partial}{\partial \theta_i} \frac{e^{\theta_k}}{\sum_j e^{\theta_j}}$$

לחלק מסתמי:

$$\frac{\partial}{\partial \theta_i} \frac{e^{\theta_i}}{\sum_j e^{\theta_j}} = \frac{e^{\theta_i} \sum_j e^{\theta_j} - e^{\theta_i} e^{\theta_i}}{(\sum_j e^{\theta_j})^2} = \frac{e^{\theta_i}}{\sum_j e^{\theta_j}} \cdot \frac{\sum_j e^{\theta_j} - e^{\theta_i}}{\sum_j e^{\theta_j}} =$$

$$= \hat{y}_i \cdot (1 - \hat{y}_i)$$

$$\frac{\partial}{\partial \theta_i} \frac{e^{\theta_k}}{\sum_j e^{\theta_j}} = \frac{-e^{\theta_k} \cdot e^{\theta_i}}{(\sum_j e^{\theta_j})^2} = - \frac{e^{\theta_k}}{\sum_j e^{\theta_j}} \cdot \frac{e^{\theta_i}}{\sum_j e^{\theta_j}} = -\hat{y}_k \hat{y}_i$$

לכאן מסתמי:

$$\frac{\partial}{\partial \theta_i} \frac{e^{\theta_k}}{\sum_j e^{\theta_j}} = \begin{cases} \hat{y}_i (1 - \hat{y}_i) & , k=i \\ -\hat{y}_k \hat{y}_i & , k \neq i \end{cases} = \hat{y}_k (\delta_{ki} - \hat{y}_i)$$

אם δ_{ki} הוא
פונקציית
קרוניקרו
אז מסתמי נקבל:

$$\dots = - \sum_k y_k \cdot \frac{1}{\hat{y}_k} \cdot \hat{y}_k (\delta_{ki} - \hat{y}_i) =$$

$$= -y_i \cdot (1 - \hat{y}_i) + \sum_{k \neq i} y_k \hat{y}_i = -y_i + y_i \hat{y}_i + \sum_{k \neq i} y_k \hat{y}_i =$$

$$= -y_i + \hat{y}_i \left(y_i + \sum_{k \neq i} y_k \right) =$$

$$= \hat{y}_i \left(\sum_k y_k \right) - y_i = \boxed{\hat{y}_i - y_i}$$

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$$\frac{\partial J}{\partial x} = \frac{\partial CE(y, \hat{y})}{\partial (xW_2 + b_2)} \cdot \frac{\partial (xW_2 + b_2)}{\partial (\sigma(xW_2 + b_2))} \cdot \frac{\partial (\sigma(xW_2 + b_2))}{\partial (xW_2 + b_2)} \quad (2)$$

$$= \frac{\partial (xW_2 + b_2)}{\partial x} = (\hat{y} - y) \cdot \frac{\partial [\sigma(xW_2 + b_2)W_2 + b_2]}{\partial (\sigma(xW_2 + b_2))} \cdot \sigma'(xW_2 + b_2) =$$

$$= (\hat{y} - y) \cdot W_2^T \cdot \sigma'(xW_2 + b_2) \cdot W_1^T =$$

$$= \boxed{(\hat{y} - y) W_2^T [\sigma(xW_2 + b_2) \cdot (1 - \sigma(xW_2 + b_2))] W_1^T}$$

$$h_{drop,i} = \begin{cases} \gamma \cdot h_i & \text{w.p. } 1 - p_{drop} \\ 0 & \text{w.p. } p_{drop} \end{cases}$$

(a) (3)

$$E[h_{drop,i}] = [1 - p_{drop}] \cdot \gamma h_i = \gamma h_i \Rightarrow$$

$$\Rightarrow \gamma = \frac{1}{1 - p_{drop}}$$