

LMevaluating perplexity



$$\begin{aligned}
PPL(\mathbf{X}^{(i)}) &= \exp \left\{ -\frac{1}{T} \sum_{t=1}^T \log P \left(\mathbf{x}^{(t)}_i \mid \mathbf{x}^{(t-n)}_i \dots \mathbf{x}^{(t-1)}_i; \boldsymbol{\theta} \right) \right\} \\
&= \exp \left\{ H(P_D, P_{\boldsymbol{\theta}}) \right\}
\end{aligned}$$

Language Learning (LLM) Leaderboard

LM evaluation using perplexity

$$\begin{aligned} PPL(\mathbf{X}^{(i)}) &= \exp \left\{ -\frac{1}{T} \sum_{t=1}^T \log P\left(\mathbf{x}^{(t)}_i \mid \mathbf{x}^{(t-n)}_i \dots \mathbf{x}^{(t-1)}_i; \boldsymbol{\theta}\right) \right\} \\ &= \exp \left\{ H(P_D, P_{\boldsymbol{\theta}}) \right\} \end{aligned}$$

Language modeling (LM) Leaderboard

Transfer learning

- Neural language modeling provides a means to extract semantically rich features from text data. How is this then applied towards downstream tasks?
- Transfer learning:
 - LMs are first trained on next-word or next-sentence prediction tasks on a very large corpus of text (e.g., wikitext 103).
 - The output layer, which is used to map onto the word prediction output space, is removed, and replaced with a randomly initialized projection matrix mapping to the output space of the particular prediction task of interest. It is trained on a much lower number of examples from that task.

