
Perplexity

Like entropy, **perplexity** provides a measure of the amount of uncertainty of a random variable. In fact, perplexity is simply a monotonic function of entropy:

$$\text{Perp}(X) := 2^{H(X)}$$

That is, the perplexity of a random variable X is simply two raised to the entropy of X .

Definition 1 *Given a random variable X , its **perplexity** is*

$$\text{Perp}(X) := 2^{H(X)}$$

where $H(X)$ is the entropy of X .

Intuition

Perplexity is often more interpretable than the entropy of a random variable and is therefore often used when we want uncertainty to be expressed in more “human” terms. The interpretation of perplexity arises from the fact that the perplexity of a uniform, discrete random variable with K outcomes is K (Theorem 1). For example, the perplexity of a fair coin is two and the perplexity of a fair six-sided die is six. This provides a frame of reference when interpreting a perplexity value. That is, if the perplexity of some random variable X is \mathcal{P} , our uncertainty towards the outcome of X is equal to the uncertainty we would have towards a \mathcal{P} -sided die.

Theorem 1 *Given a discrete uniform random variable*

$$X \sim \text{Cat}(p_1, p_2, \dots, p_K)$$

where $\forall i, j \in [K] \ p_i = p_j$, the perplexity of X is K .

Proof:

$$\begin{aligned}\text{Perp}(X) &= 2^{H(X)} \\ &= 2^{-\frac{1}{K} \sum_{i=1}^K \log_2 \frac{1}{K}} \\ &= 2^{-\log_2 \frac{1}{K}} \\ &= \frac{1}{2^{\log_2 \frac{1}{K}}} \\ &= \frac{1}{1/K} \\ &= K\end{aligned}$$

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