## **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:

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

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

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

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 provide 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

$$X \sim Cat(p_1, p_2, \ldots, p_K)$$

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

**Proof:** 

Perp(X) = 
$$2^{H(X)}$$
  
=  $2^{-\frac{1}{K}\log_2\frac{1}{K}}$   
=  $2^{-\log_2\frac{1}{K}}$   
=  $\frac{1}{2^{\log_2\frac{1}{K}}}$   
=  $\frac{1}{1/K}$   
=  $K$