

Instead of calling your $\Delta(w, t)$ Dirac delta, it is better to call it Heaviside function $H\left(x - \frac{1}{2}\right) = \frac{1}{2}(1 + \operatorname{sgn}(x - \frac{1}{2}))$ https://en.wikipedia.org/wiki/Heaviside_step_function

Good thing about Heaviside function is that it can be approximated by the smooth logistic function $H(x) = \frac{1}{1+e^{-2kx}}$, with k large enough (along with other smooth approximations). This may both be related to entropies and used as a differentiable loss function in optimizing ML models (not sure how to at the moment).

$N_p(w, *) = \sum_t N^p(w, t)$ technically is not a norm, as it does not satisfy homogeneity property (except for $p = 1$). It is a pseudonorm though. Similarly to the above $\Delta(w, *)$ is a pseudonorm and is Hamming distance of such vector from zero.

It may be useful to consider a vector space over vocabulary $T = \{t\}$ of size $|T|$ and a vector N in this space with $|T|$ components $N(w, t)$. It is interesting to note that an entropy-like expression can be obtained for actual l_1 norm, because

$$\begin{aligned} \frac{d}{dp} \|N\|_{l_p} &= \frac{d}{dp} \left(\sum_t N^p(w, t) \right)^{1/p} = \log \sum_t N^p(w, t) \left(\sum_t N^p(w, t) \right)^{1/p} \sum_t \log N(w, t) N^p(w, t) = \\ &= H(w, *) \|N\|_{l_p} \log \sum_t N^p(w, t) \end{aligned}$$

I don't have a viable physical interpretation for this, but it may be highly far-reaching.