Basic idea to reformulate Recitfied Linear Unit (ReLu)

$$ReLu(x) = ln(1 + exp(x))$$

In an MLP, from one layer to the next, each node activation  $\mathbf{z} = \mathbf{W}\mathbf{h} + \mathbf{b}$  WLOG, let the bias  $\mathbf{b} = 0$ . Also, assume ReLu is applied elementwise to the vector  $\mathbf{z}$ 

Then, for each activated node  $\mathbf{n}_i$ 

$$\begin{split} n_i &= ReLu(\mathbf{z}_j) = ln(1 + exp((\mathbf{W}\mathbf{h})_j) \\ &= ln(1 + exp((\sum_i \mathbf{W_{ij}h_i})) \\ &= ln(1 + \prod_i exp((\mathbf{W_{ij}h_i})) \end{split}$$

Which seems intractable

But

What if want consider  $Relu(\mathbf{z}_j)$  in Expectation

$$\mathbb{E}[Relu(\mathbf{z}_i)]$$

we assume that each activated node exists in a superposition of 2 states [0,1] what I want to do is somehow get the superposition inside the log, so that

$$\mathbb{E}[Relu(\mathbf{z}_i)] \approx ln(\mathbb{E}[1 + exp((\mathbf{Wh})_i])$$

this is kind of like the annealed approximation in spin glass theory—which is not a great approximation, but perhaps, can give some initial insight that is

$$\mathbb{E}[ln\mathbf{Z}] \approx ln(\mathbb{E}[\mathbf{Z}])$$

which is not generally true for spin glasses, and why we use replica theory, cavity method, etc

Also, lets introduce an internal state vector  $\mathbf{c} = [0, 1]$ 

$$ln(\mathbb{E}[1 + exp((\mathbf{Wh})_j]) = ln(\mathbb{E}[exp((\mathbf{c^TWh})_j])$$

and we can get something like

$$\mathbb{E}[Relu(\mathbf{z}_j)] \approx ln(\mathbb{E}[exp((\mathbf{c^TWh})_j])$$

We now pull the expecation back out...

$$\approx \mathbb{E}[ln(exp((\mathbf{c^TWh})_j))]$$

$$\approx \mathbb{E}[(\mathbf{c^TWh})_j]$$

So obviously this is very sloppy, but it is an attempt to get at defining some kind of (linearized) free energy for each layer