Notes on Tropical Nets

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1 Definitions

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Basic Setting
Inputs and Input Space x \in \mathcal{X} = \mathbb{R}^{n_0}
Network \mathcal{N}: \mathcal{X} \mapsto \mathbb{R}
Layer Indicies l \in \{0\} \cup \{1, \dots, d\} \cup \{d+1\} (0 for input; d+1 to output)
Hidden Layer Widths n_1, \ldots, n_d.
Total neurons n = \sum_{l=1}^d n_l Weight matrices W^l \in \mathbb{R}^{n_{l+1} \times n_l} for l < d+1.
Biases b^l \in \mathbb{R}^{n_l} for l > 0.
Nonlinearity R(x)_i = \max(0, x_i)
Network definition:
                                      \mathcal{N}^{l+1}(x) = b^{l+1} + W^l R(\mathcal{N}^l(x)) \qquad \mathcal{N}(x) = \mathcal{N}^{d+1}(x)
\mathcal{N}^1(x) = b^1 + W^0 x
States and Linear Regions
Layer State Map \sigma^l: \mathcal{X} \mapsto \{0,1\}^{n_l} where each coordinate, \sigma^l(x)_i, is an indicator for
\mathcal{N}^l(x)_i > 0. ("Neuron State")
Network State Map \sigma(x) = (\sigma^d(x), \dots, \sigma^1(x))
Network States \Sigma = {\sigma(x) : x \in \mathcal{X}}
As a shorthand, identify \Sigma \subset \{0,1\}^n with flattened counterparts
For any \mu = (\mu^d, \dots, \mu^1) \in \{0, 1\}^n:
Identify each \mu^l appearing in linear equations with diagonal matrix, diag(\mu^l)
           for example: R(\mathcal{N}^l(x)) = \sigma^l(x)\mathcal{N}^l(x)
Corresp. Linear Function: L_{\mu} by replacing layer l with multilpication by \mu^{l}.
Corresp. Region: \mathcal{X}_{\mu} = \overline{\{x \in \mathcal{X} : \sigma(x) = \mu\}}
            so that: \mathcal{X}_{\mu} \subset \{x \in \mathcal{X} : \mathcal{N}(x) = L_{\mu} \left( \frac{x}{1} \right) \} (Corresp. Linear Region)
Tropical Notation
For any vector b or matrix W,
define b_{\pm}=R(\pm b) (W_{\pm})_{i,j}=R(\pm W_{i,j}) such that b=b_{+}-b_{-} W=W_{+}-W_{-} likewise define |b|=b_{+}+b_{-} |W|=W_{+}+W_{-} Define recursively for x\in\mathcal{X} \mu,\tau\in\{0,1\}^{n}:
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Table 1: Example First Few Recursions (without biases)

layer l	$f^l_{\mu}(x)$	$g^l_\mu(x)$
1	W^0x	0
2	$W^{1}_{+}\mu^{1}W^{0}x$	$W_{+}^{1}\mu^{1}W^{0}x$
3	$W_{+}^{2}\mu^{2}W^{1}\mu^{1}W^{0}x + W^{2} W_{-}^{1}\mu^{1}W^{0}x$	$ W_{-}^{1}\mu^{1}W^{0}x \\ W_{-}^{2}\mu^{2}W^{1}\mu^{1}W^{0}x + W^{2} W_{-}^{1}\mu^{1}W^{0}x $

$$\begin{array}{c|ccccc} \text{layer } l & p_{\mu}^{l}(x) & q_{\mu}^{l}(x) \\ \hline 1 & \mu^{l}W^{0}x & W_{-}^{0}x \\ 2 & \mu^{2}W^{1}\mu^{1}W^{0} & W_{-}^{1}\mu^{1}W^{0}x + |W^{1}|W_{-}^{0}x \\ 3 & \mu^{3}W^{2}\mu^{2}W^{1}\mu^{1}W^{0}x & W_{-}^{2}\mu^{2}W^{1}\mu^{1}W^{0}x + |W^{2}|(W_{-}^{1}\mu^{1}W^{0}x + |W^{1}|W_{-}^{0}x) \end{array}$$

Recursion 1

$$\begin{array}{l} \frac{Recursion 1}{f_{\mu}^{1}(x) = W^{0}x} & g_{\mu}^{1}(x) = 0 \\ f_{\mu}^{l+1}(x) = b^{l+1} + W_{+}^{l}\mu^{l}(f_{\mu}^{l}(x) - g_{\mu}^{l}(x)) + |W^{l}|g_{\mu}^{l}(x) \\ g_{\mu}^{l+1}(x) = b^{l+1} + W_{-}^{l}\mu^{l}(f_{\mu}^{l}(x) - g_{\mu}^{l}(x)) + |W^{l}|g_{\mu}^{l}(x) \\ f_{\mu}(x) = f_{\mu}^{d+1}(x) & g_{\mu}(x) = g_{\mu}^{d+1}(x) \\ \frac{Recursion 2}{p_{\mu}^{0}(x) = x} & q_{\mu}^{0}(x) = 0 \\ p_{\mu}^{l+1}(x) = \mu^{l+1}(b^{l+1} + W^{l+1}p_{\mu}^{l}(x)) \\ q_{\mu}^{l+1}(x) = W_{-}^{l}p_{\mu}^{l}(x) + |W^{l}|q_{\mu}^{l}(x) \\ \frac{Relation to Network}{N^{l}(x) = f_{\sigma(x)}^{l}(x) - g_{\sigma(x)}^{l}(x) \text{ for } l = 1, \dots, d+1 \\ R(\mathcal{N}^{l}(x)) = p_{\sigma(x)}^{l}(x) \end{array}$$

Theorem 1. Tropical Formulation

For $l = 0, \ldots, d$

$$\mathcal{N}^{l+1}(x) = \left(\max_{\mu \in \{0,1\}^n} f_{\mu}^{l+1}(x)\right) - \left(\max_{\tau \in \{0,1\}^n} g_{\tau}^{l+1}(x)\right) \tag{1}$$

$$R(\mathcal{N}^l)(x) = \left(\max_{\mu \in \{0,1\}^n} p_{\mu}^l(x) + q_{\mu}^l(x)\right) - \left(\max_{\tau \in \{0,1\}^n} q_{\tau}^l(x)\right) \tag{2}$$

where each term is an affine function of x.