

## Assignment 2: Expressiveness

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### Part 1: Boolean AND-OR Networks

1. We will show that the lower bound on  $B$  cannot be larger than  $2^{d-1}$ , since  $XOR_d$  is realizable by a shallow network of width  $B = 2^{d-1}$ .

Denote the truth table of  $XOR_d$ :

$x_1$	$x_2$	$\dots$	$x_d$
+1	-1	$\dots$	-1
$\dots$	$\dots$	$\dots$	$\dots$
+1	+1	$\dots$	+1

Denote each row  $i = (x_1, \dots, x_d)$  where  $XOR_d(x_1, \dots, x_d) = +1$  as  $w_i$ .

Notice that:

- a.  $\forall v \in \{+1, -1\}^d. v * v = (+1)^d$
- b.  $\forall u, v \in \{+1, -1\}^d \exists k. (u * v)_k = -1$

We define the following shallow network, mark as  $N$ :

- $B = 2^{d-1}$
- $U = I$
- $W = \begin{pmatrix} -w_1 & - \\ \dots & \\ -w_{2^{d-1}} & - \end{pmatrix}$

We will show that:

1.  $\forall (x_1, \dots, x_d) \text{ s.t. } XOR_d(x_1, \dots, x_d) = +1. N(x_1, \dots, x_d) = +1$
2.  $\forall (x_1, \dots, x_d) \text{ s.t. } XOR_d(x_1, \dots, x_d) = -1. N(x_1, \dots, x_d) = -1$

1. Let  $x_1, \dots, x_d$  be inputs  $\in \{+1, -1\}^d$  s.t  $XOR_d(x_1, \dots, x_d) = +1$ , so  $(x_1, \dots, x_d) = w_i$  for some  $1 \leq i \leq 2^{d-1}$ . We'll show that  $N(x_1, \dots, x_d) = +1$ .

$$N(x) = OR(U * AND(W * x))$$

$$AND(W * x) = AND\left(\begin{pmatrix} -w_1 & - \\ \dots & \\ -w_{2^{d-1}} & - \end{pmatrix} \begin{pmatrix} | \\ w_i \\ | \end{pmatrix}\right) = (-1, -1, \dots, +1, \dots, -1)$$

Where the only +1 is in index  $i$  (by claim a) .

$$OR(U * (-1, -1, \dots, +1, \dots, -1)) = OR((-1, -1, \dots, +1, \dots, -1)) = +1 \blacksquare$$

2. Similarly, let  $x_1, \dots, x_d$  be inputs  $\in \{+1, -1\}^d$  s.t  $XOR_d(x_1, \dots, x_d) = -1$ . So,  $\forall i. x \neq w_i$ .

$$\text{Therefore, } AND(W * x) = AND\left(\begin{pmatrix} -w_1 & - \\ \dots & \\ -w_{2^{d-1}} & - \end{pmatrix} \begin{pmatrix} 1 \\ x \\ 1 \end{pmatrix}\right) = (-1)^d \rightarrow OR(U * (-1)^d) = -1 \blacksquare$$

2. Let us look at the number of functions a deep network can express:

- $2 \log_2(d)$  Layers:
  - Odd layers:
    - $W \in \mathbb{R}^{\bar{B} * d}$
    - $W^{(2)}, \dots, W^{(2 \log_2(d))} \in \mathbb{R}^{\bar{B} * \bar{B}}$
    - $\forall k. W_{ij}^{(k)} \in \{-1, 0, +1\}$
  - Even layers:
    - $U^{(1)}, \dots, U^{(2 \log_2(d)-1)} \in \mathbb{R}^{\bar{B} * \bar{B}}$
    - $U^{(2 \log_2(d))} \in \mathbb{R}^{\bar{B} * 1}$
    - $\forall k. U_{ij}^{(k)} \in \{-1, +1\}$

Mark T as the total number of parameters in the network:

$$T = (2 \log_2(d) - 2) * (\bar{B} * \bar{B}) + (\bar{B} * d) + (\bar{B} * 1)$$

The number of possible functions is therefore  $2^{T+C} = 2^{\text{poly}(\bar{B})}$  (where C is some constant to change the exponent base from 3 to 2).

$|Y^X| = |Y|^{|X|} = 2^{2^d} \rightarrow \text{unless } \bar{B} \in \exp(d), 2^{\text{poly}(\bar{B})} < 2^{2^d}$ , so there exists a function that the network will not be able to express.

## Part 2: Fully Connected ReLU Networks with 1D Input

1. First, we note that  $B \geq 2$  is required for the condition to hold: when  $B = 1$ ,  $h \in H_1$  is equal to  $h(x) = w_2[w_1x + b_1]_+ + b_2$ , this is a PWL function with 2 pieces- a constant part (negative values after Relu) and a linear part. Therefore, the number of pieces  $= 2 \not\leq 1 = B$ .

**A shallow network of width  $B \geq 2$  can realize any PWL mapping with  $\leq B$  pieces:**

For  $B \geq 2$  we will prove by induction by explicitly building such  $h$ .

$$h(x) = \sum_{i=1}^B w_2^i [w_1^i x + b_1^i]_+ + b_2$$

with  $g_1(x) = w_2^1 [w_1^1 x + b_1^1]_+ + b_2$  and  $g_i(x) = w_2^i [w_1^i x + b_1^i]_+$

So  $h(x) = \sum_{i=1}^B g_i(x)$

**Base case B=2:**

Let  $f$  be a PWL function with  $B=2$  pieces, denote  $f(x) = \begin{cases} m_1x + n_1 & x < p \\ m_2x + n_2 & x \geq p \end{cases}$

We will define two functions  $g_1, g_2: \mathbb{R} \rightarrow \mathbb{R}$ , and show that  $h(x) := (g_1 + g_2)(x) \equiv f(x)$ :

Set  $w_1^1 = -1, w_2^1 = -m_1, b_1^1 = p, b_2 = n_1 + m_1p$  and get:

$$g_1(x) = -m_1[p - x]_+ + n_1 + m_1p = \begin{cases} m_1x + n_1 & x < p \\ m_1p + n_1 & x \geq p \end{cases}$$

Now set  $w_1^2 = 1, w_2^2 = m_2, b_1^2 = -p$  and get:

$$g_2(x) = m_2[x - p]_+ = \begin{cases} 0 & x < p \\ m_2x - m_2p & x \geq p \end{cases}$$

Overall, we get:

$$\begin{aligned} (g_1 + g_2)(x) &= \begin{cases} m_1x + n_1 & x < p \\ m_1p + n_1 + m_2x - m_2p & x \geq p \end{cases} \\ &= \begin{cases} m_1x + n_1 & x < p \\ m_1p + n_1 + m_2x - m_2p & x \geq p \end{cases} \stackrel{(*)}{=} \begin{cases} m_1x + n_1 & x < p \\ m_2x + n_2 & x \geq p \end{cases} = f(x) \end{aligned}$$

(\*)  $f$  is continuous so  $m_1p + n_1 = m_2p + n_2$

**Induction Step:**

Let's assume there exist some  $B$  so that for any PWL  $f \exists h \in H_B$  s.t.  $h(x) = f(x)$

Let  $f$  be a PWL function with  $B+1$  pieces  $\{(-\infty, p_1), [p_1, p_2), \dots, [p_B, \infty)\}$

$f(x) = m_i x + n_i$  if  $x \in [p_{i-1}, p_i)$  for  $i \in [B+1]$  (denote  $p_0 = -\infty, p_{B+1} = \infty$ )

Let  $f^*(x) = m_i x + n_i$  if  $x \in [p_{i-1}, p_i)$  for  $i \in [B]$

By induction we know that there exists  $h^* \in H_B$  s.t.  $h^*(x) = f^*(x)$ , with  $h^*(x) = \sum_{i=1}^B g_i(x)$

And define  $g_{B+1}(x) = w_2^{B+1} [w_1^{B+1} x + b_1^{B+1}]_+$

with  $w_1^{B+1} = 1, w_2^{B+1} = m_{B+1} - m_B, b_1^{B+1} = -p_B$

So overall we will get:

$$\begin{aligned}
h^*(x) + g_{B+1}(x) &= \begin{cases} h^*(x) & x < p_B \\ m_B x + n_B + (m_{B+1} - m_B)(x - p_B) & x \geq p_B \end{cases} \\
&= \begin{cases} h^*(x) & x < p_B \\ m_B x + n_B + m_{B+1}x - m_B x + m_{B+1}p_B - m_B p_B & x \geq p_B \end{cases} \stackrel{(*)}{=} \\
\begin{cases} h^*(x) & x < p \\ m_{B+1}x + m_{B+1}p_B + n_{B+1} - m_{B+1}p_B & x \geq p \end{cases} &= \begin{cases} h^*(x) & x < p \\ m_{B+1}x + n_{B+1} & x \geq p \end{cases} = f(x)
\end{aligned}$$

(\*) again,  $f$  is continuous so  $m_i p_i + n_i = m_{i+1} p_i + n_i$  ■

Any mapping realizable by such network is PWL with  $\leq B + 1$  pieces:

Let  $h \in H_B$ , so  $h$  is PWL with  $\leq B + 1$  pieces because  $h$  can be written as a sum of PWL functions (like in the first direction):

Denote  $p = \arg_x [w_1 \cdot x + b_1 = 0]$ , let  $\text{argsort}(p) = (i_1, i_2, \dots, i_B)$  so that  $p_{i_1} \leq p_{i_2} \leq \dots \leq p_{i_B}$

Let  $g_{i_1}(x) = w_2^{i_1} [w_1^{i_1} x + b_1^{i_1}]_+ + b_2$  and for  $i_1 \neq i \in \text{argsort}(p)$ ,  $g_i(x) = w_2^i [w_1^i x + b_1^i]_+$

$$h(x) = \sum_{j \in \text{argsort}(p)} g_j(x) = \begin{cases} g_{i_1}(x) & x < p_{i_1} \\ g_{i_1}(x) + g_{i_2}(x) & p_{i_1} \leq x < p_{i_2} \\ \vdots & \\ \sum_{j \in \text{argsort}(p)} g_j(x) & x \geq p_B \end{cases}$$

all pieces of  $h$  are linear functions as sum of linear functions, and so by the equation above we have  $\leq B + 1$  pieces of different linear functions. ■

2. As shown in class,

$$|S_{<}| = \# \text{ of sawteeth} = 2^{L-2}, \quad |S_{>}| = 2^{L-2} + 1 \Rightarrow |S_{<} \cup S_{>}| = 2^{L-1} + 1$$

We'll show that a PWL function  $h$  with  $\leq B + 1$  pieces could avoid missing  $\leq$

$$\left\lceil \frac{1}{2}(2^{L-1} + 1) + \frac{1}{2}(B + 1) \right\rceil \text{ out of } 2^{L-1} + 1 \text{ intervals in } S_{<}, S_{>}$$

**Claim:** linear  $h$  ( $B=1$ ) can avoid missing  $\leq \left\lceil \frac{1}{2}(2^{L-1} + 1) + \frac{1}{2} \right\rceil$

$h$  is linear (assume WLOG  $h$  increasing)  $\Rightarrow h$  intersect with  $y = \frac{1}{2}$  once, denote the intersection point by  $x^*$  (assume WLOG  $x^*$  is a shared point of  $S_{>}$  and  $S_{<}$ , that is there exists  $[x_1, x^*] \in S_{<}$  and  $[x^*, x_2] \in S_{>}) \Rightarrow h$  is below  $y = \frac{1}{2}$  for all  $x < x^*$ , and above it for all  $x > x^*$ . Denote the number of intervals  $\in S_{<} \cup S_{>}$  in  $[0, x^*]$  by  $k_1$  and the number of intervals  $\in S_{<} \cup S_{>}$  in  $[x^*, 1]$  by  $k_2$ , note that  $k_1 + k_2 = |S_{<} \cup S_{>}| = 2^{L-1} + 1$ , so  $h$  avoids missing no more than half of  $k_1$  and no more than half of  $k_2$ , that is  $\leq \frac{1}{2}k_2 + \frac{1}{2}k_1 = \frac{1}{2}(2^{L-1} + 1) \leq \left\lceil \frac{1}{2}(2^{L-1} + 1) + \frac{1}{2} \right\rceil$

back to a general PWL  $h$  with  $\leq B + 1$  pieces:

Let's divide  $[0,1]$  into the intervals  $[0 = C_0, C_1], [C_1, C_2], \dots, [C_B, 1 = C_{B+1}]$  where on each interval  $h$  is linear. Denote the number of *intervals*  $\in S_{<} \cup S_{>}$  in  $(C_i, C_{i+1})$  by  $k_{i+1}$  so we get  $\sum_i k_i = |S_{<} \cup S_{>}| = 2^{L-1} + 1$ .

Now for each  $[C_i, C_{i+1}]$   $h$  is linear and the claim above applies, so over all we get that the number of intervals  $h$  can avoid missing is  $\leq \sum_i \left[ \frac{1}{2}(k_i) + \frac{1}{2} \right] = \left[ \frac{1}{2}(2^{L-1} + 1) + \frac{1}{2}(B + 1) \right]$  ■

3. The universality and expressive efficiency analyses given in class, with the following modifications applies to leaky ReLU activation  $\sigma(\mathbf{z}) = \max\{\mathbf{a}\mathbf{z}, \mathbf{z}\}$ ,  $\mathbf{a} \in (0, 1)$ :

**Universality:** for any  $f \in \mathcal{F}$ ,  $\epsilon > 0$  there exist  $h \in \mathcal{H}_B$  that is built with leaky ReLU such that  $d(f, h) < \epsilon$ .

Note that this will also stand for  $h \in \bar{\mathcal{H}}_B$  as  $\mathcal{H}_B \subseteq \bar{\mathcal{H}}_B$ .

We have seen in class that we can approximate any continuous function with PWL one, and that any PWL function can be realized by a ReLU network. In addition, we have seen that any ReLU neuron can be implemented by 2 leaky ReLU neurons, and vice versa.

So overall, any  $f$  can be approximated by PWL  $h$  with  $B$  pieces, that can be realized by a shallow ReLU network of width  $B$ , that can be modified to a leaky ReLU network of width  $2B$ .

**Expressive efficiency:**

$\forall \bar{B} \in O(B)$  s.t.  $\mathcal{H}_B \subseteq \bar{\mathcal{H}}_{\bar{B}}$  where here  $\mathcal{H}_B$  is the set of functions that can be realized by leaky ReLU neural network.

$\exists \bar{h} \in \bar{\mathcal{H}}_{\bar{B}}$  for  $\bar{B} \in O(1)$  s.t.  $\bar{h} \notin \mathcal{H}_B$  unless  $B \in \exp(L)$

As seen in class, condition (i) is an immediate consequence of the fact that a deep network can realize any mapping realizable by a shallow one of the same width.

Now for condition (ii) we'll construct a sawteeth function  $g := g^{eL}$  similar to the one from class, now  $g$  is PWL with  $2^{L-1}$  pieces and we've seen that it can be realized by an  $L$  layer ReLU NN of width 3, which means the same function can be realized by an  $L$  layer leaky ReLU NN of width 6 (  $*$  ) as any ReLU neuron can be implemented by 2 leaky ones and vice versa).

Next, let  $h \in \mathcal{H}_B$  be a function that can be realized by a shallow leaky ReLU NN of width  $B \Rightarrow h$  is PWL with  $\leq 2B + 2$  (again because  $*$ ) pieces and like the proposition in class the most intervals in  $S_{<} \cup S_{>}$   $h$  can avoid missing is  $\leq \left[ \frac{1}{2}(2^{L-1} + 1) + \frac{1}{2}(2B + 2) \right]$  and like we've seen in class this implies that the closest a shallow leaky ReLU NN of width  $2B+2$  can get to  $g$  is  $\geq \frac{1}{4} - (2B)2^{-L-1} - 3 \cdot 2^{-L}$  requiring  $B > 2^L \left( \frac{1}{4} - \epsilon \right) - 6 \in \exp(L)$  ■

4. let  $f$  be a Reimann integrable function in  $[0,1]$ ,  $\epsilon^* > 0$ ,  $B \in \mathbb{N}$ , denote  $\epsilon = \frac{\epsilon^*}{B}$ .

$f$  is integrable so there exist  $T, S: \mathbb{R} \rightarrow \mathbb{R}$  and intervals  $(x_i, x_{i+1}] \forall i \in [B]$  in which  $S, T$  are constant functions (step functions) s.t.  $\forall x \in [0,1]$ ,  $S(x) \leq f(x) \leq T(x)$  and

$$T(x) - S(x) < \epsilon$$

define  $h: \mathbb{R} \rightarrow \mathbb{R} \forall x \in (x_i, x_{i+1}]$  by  $h(x) = x \frac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i} + x_i$  that is the line connecting the points  $(x_i, f(x_i)), (x_{i+1}, f(x_{i+1}))$ .

Note that by definition  $h$  is PWL with  $B$  pieces and so  $h \in \mathcal{H}_B \subseteq \bar{\mathcal{H}}_B$ .

Now, we know that  $\forall x \in (x_i, x_{i+1}] \max_x |f(x)| \leq T(x_i)$  and  $\min_x |h(x)| \geq S(x_i)$  so we get

$$\begin{aligned} d(f, h) &= \int_0^1 |f(x) - h(x)| dx \leq \int_0^1 |f(x)| dx - \int_0^1 |h(x)| dx \leq \int_0^1 |T(x)| dx - \int_0^1 |S(x)| dx \\ &= \sum_i T(x_i)(x_{i+1} - x_i) - \sum_i S(x_i)(x_{i+1} - x_i) \\ &= \sum_i (T(x_i) - S(x_i))(x_{i+1} - x_i) \\ &\stackrel{\forall x, T(x) - S(x) < \epsilon}{\lesssim} \sum_i \epsilon(x_{i+1} - x_i) \stackrel{\sum_i (x_{i+1} - x_i) = 1}{\cong} B\epsilon \sum_i (x_{i+1} - x_i) = B\epsilon = \epsilon^* \quad \blacksquare \end{aligned}$$

### Part 3: Convolutional Arithmetic Circuits

1. Let  $T$  and  $\bar{T}$  be tensors of orders  $n$  and  $\bar{n}$  respectively. Let  $I \subset [n + \bar{n}]$  and denote by  $I - n$  the set obtained by subtracting  $n$  from each element of  $I$ , Then:  $\llbracket T \otimes \bar{T} \rrbracket_I = \llbracket T \rrbracket_{I \cap [n]} \odot \llbracket \bar{T} \rrbracket_{I - n \cap [\bar{n}]} \cdot$

WLOG we can assume the partition is over the first indices, otherwise we can apply a permutation for each of the tensors. By definition we get:

$$\begin{aligned} \llbracket T \rrbracket_{I \cap [n]} \odot \llbracket \bar{T} \rrbracket_{I - n \cap [\bar{n}]} &= \llbracket T \rrbracket_{I \cap [n]} * \llbracket \bar{T} \rrbracket_{I - n \cap [\bar{n}]} \\ &= [T]_{i_1, j_1} * [\bar{T}]_{i_2, j_2} = [T \otimes \bar{T}]_{i_1, j_1, i_2, j_2} \\ &= \llbracket T \otimes \bar{T} \rrbracket_{I \cap [n] * (i_1 + j_1), I - n \cap [\bar{n}] * (i_2 + j_2)} \end{aligned}$$

This holds for every  $i_1, j_1, i_2, j_2$  where  $i_1 \in I \cap [n], j_1 \in I^c \cap [n], i_2 \in I - n \cap [\bar{n}], j_2 \in I^c - n \cap [\bar{n}]$ , and thus  $\llbracket T \rrbracket_{I \cap [n]} \odot \llbracket \bar{T} \rrbracket_{I - n \cap [\bar{n}]} = \llbracket T \otimes \bar{T} \rrbracket_I$  ■

2. Modifications are only needed for part (ii) of the proof shown in class:

We apply canonical matricization to the HT decomposition (corresponding to the deep network), while using the linearity of matricization and its relationship with outer/Kronecker product:

$$\llbracket \phi^{1,j,\gamma} \rrbracket = \sum_{\alpha=1}^{r_0} a_{\alpha}^{1,j,\gamma} \llbracket a^{0,4j-1,\alpha} \otimes a^{0,4j,\alpha} \rrbracket, \quad \text{for } j = 1, 2, \dots, \frac{N}{4}, \gamma = 1, 2, \dots, r_1$$

$$\llbracket \phi^{l,j,\gamma} \rrbracket = \sum_{\alpha=1}^{r_{l-1}} a_{\alpha}^{l,j,\gamma} \llbracket \phi^{l-1,4j-1,\alpha} \rrbracket \odot \llbracket \phi^{l-1,4j,\alpha} \rrbracket, \quad \text{for } j = 1, 2, \dots, \frac{N}{4^l}, \gamma = 1, 2, \dots, r_l$$

$$\llbracket \phi^{L-1,j,\gamma} \rrbracket = \sum_{\alpha=1}^{r_{L-2}} a_{\alpha}^{L-1,j,\gamma} \llbracket \phi^{L-2,4j-1,\alpha} \rrbracket \odot \llbracket \phi^{L-2,4j,\alpha} \rrbracket, \quad \text{for } j = 1, 2, 3, 4, \gamma = 1, 2, \dots, r_{L-1}$$

$$\llbracket A^{HT} \rrbracket = \sum_{\alpha=1}^{r_{L-1}} a_{\alpha}^L \llbracket \phi^{L-1,1,\alpha} \rrbracket \odot \llbracket \phi^{L-1,2,\alpha} \rrbracket \odot \llbracket \phi^{L-1,3,\alpha} \rrbracket \odot \llbracket \phi^{L-1,4,\alpha} \rrbracket$$

Let  $r_0$  be  $\geq M$ , and assign its filter sets -  $\{\underline{a}^{0,1,\gamma}, \dots, \underline{a}^{0,N,\gamma}\}_{\gamma=1}^{r_0}$  by  $\underline{a}^{0,j,\gamma} = e^{\gamma}$  for  $\gamma \in [M]$  and  $\underline{a}^{0,j,\gamma} = \underline{0}$  for  $\gamma > M$ .

Assign the filter sets in hidden layer 1 -  $\{\underline{a}^{1,1,\gamma}, \dots, \underline{a}^{1,N/4,\gamma}\}_{\gamma=1}^{r_1}$  - by  $\underline{a}^{1,j,1} = \underline{1}$  (all ones) and  $\underline{a}^{1,j,\gamma} = \underline{0}$  for  $\gamma > 1$ .

Assign the filter sets in hidden layers 2 through L-1 by:

$$\underline{a}^{l,j,\gamma} = \begin{cases} \underline{e}^1 & \text{for } \gamma = 1 \\ \underline{0} & \text{otherwise} \end{cases}$$

where  $l = 2, 3, \dots, L-1$ ,  $j = 1, 2, \dots, \frac{N}{4^l}$ ,  $\gamma = 1, 2, \dots, r_l$ .

And the output weights by  $\underline{a}^L = \underline{e}^1$ . Under our assignments:

$$\llbracket A^{HT} \rrbracket = Id_m \underbrace{\bigodot \dots \bigodot}_{\frac{N}{4} \text{ times}} Id_m = Id_{\frac{N}{M^4}}$$

In particular,  $rank(\llbracket A^{HT} \rrbracket) = M^{\frac{N}{4}}$ . Our results on  $rank(\llbracket A^{HT} \rrbracket)$  implies that for  $\bar{B} \geq M$ ,  $\bar{H}_{\bar{B}}$  includes a function  $\bar{h}$  whose corresponding tensor has (canonical) matricization rank  $M^{N/4}$ . On the other hand, our upper bound on  $rank(\llbracket A^{CP} \rrbracket)$  implies that  $H_B$  includes only functions whose corresponding tensors have (canonical) matricization ranks  $\leq B$ . This means that  $\bar{h} \notin H_B$  unless  $B \geq M^{\frac{N}{4}}$ , establishing condition (ii).

3. To prove  $Sep(f_{deep}; I_{quad}) \leq r_{L-1} r_{L-2}^2$ , we can use the following lemma from class:

$$Sep(f; I) = rank(A)_I$$

Where A is the tensor that identifies with the function  $f$ .

In the case of a deep network  $A = A^{HT} = \sum_{\alpha=1}^{r_{L-1}} a_{\alpha}^L * (\phi^{L-1,1,\alpha} \otimes \phi^{L-1,2,\alpha})$  and

$$\phi^{L-1,j,\gamma} = \sum_{\alpha=1}^{r_{L-2}} a_{\alpha}^{L-1,j,\gamma} \phi^{L-2,2j-1,\alpha} \otimes \phi^{L-2,2j,\alpha}, \quad \text{for } j = 1, 2, 3, 4, \gamma = 1, 2, \dots, r_{L-1}$$

Applying the matricization we get:

$$\begin{aligned} \llbracket A^{HT} \rrbracket_{I_{quad}} &= \sum_{\alpha=1}^{r_{L-1}} a_{\alpha}^L * \llbracket \phi^{L-1,1,\alpha} \rrbracket_{I_{quad}} \odot \llbracket \phi^{L-1,2,\alpha} \rrbracket_{I_{quad}} \\ &= \sum_{\alpha=1}^{r_{L-1}} a_{\alpha}^L \left[ \left[ \sum_{\alpha=1}^{r_{L-2}} a_{\alpha}^{L-1,1,\gamma} \phi^{L-2,1,\alpha} \otimes \phi^{L-2,2,\alpha}, \right]_{I_{quad} \cap \left[ \frac{N}{2} \right]} \odot \left[ \sum_{\alpha=1}^{r_{L-2}} a_{\alpha}^{L-1,2,\gamma} \phi^{L-2,3,\alpha} \otimes \phi^{L-2,4,\alpha}, \right]_{I_{quad} - \frac{N}{2} \cap \left[ \frac{N}{2} \right]} \right] \\ &= \sum_{\alpha=1}^{r_{L-1}} a_{\alpha}^L \sum_{\alpha=1}^{r_{L-2}} a_{\alpha}^{L-1,1,\gamma} \overbrace{\llbracket \phi^{L-2,1,\alpha} \rrbracket_{I_{quad} \cap \left[ \frac{N}{4} \right]}}^{\text{rank}=1} \odot \underbrace{\llbracket \phi^{L-2,2,\alpha} \rrbracket_{(I_{quad} \cap \left[ \frac{N}{2} \right]) - \frac{N}{4} \cap \left[ \frac{N}{4} \right]}}_{=\emptyset} \\ &\quad \odot \sum_{\alpha=1}^{r_{L-2}} a_{\alpha}^{L-1,2,\gamma} \underbrace{\llbracket \phi^{L-2,3,\alpha} \rrbracket_{(I_{quad} - \frac{N}{2} \cap \left[ \frac{N}{2} \right]) \cap \left[ \frac{N}{4} \right]}}_{\text{rank}=1} \odot \underbrace{\llbracket \phi^{L-2,4,\alpha} \rrbracket_{(I_{quad} - \frac{N}{2} \cap \left[ \frac{N}{2} \right]) - \frac{N}{4} \cap \left[ \frac{N}{4} \right]}}_{=\emptyset} \end{aligned}$$



Using  $rank(A \odot B) = rank(A) * rank(B)$  we get exactly  $rank\left(\llbracket A^{HT} \rrbracket_{I_{quad}}\right) \leq r_{L-1}r_{L-2}^2$