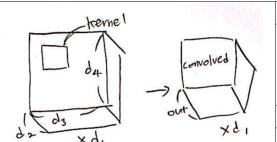
# **Matmul Layers**

## torch.nn.Linear

torch.nn.Linear(in\_features, out\_features, bias=True)(x)



## Require

- $|x| = (d_1, d_2, \dots, d_k)$
- $rank(|x|) \ge 1$
- $d_k = in\_features$

### Guarantees

•  $|y| = (d_1, d_2, \dots, d_{k-1}, out\_features)$ 

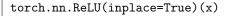
#### Comment

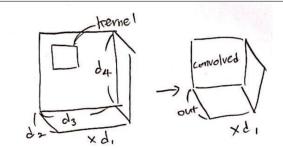
- $y = xA^T + b$ 를 계산하는 dense 레이어
- 1차원인 경우에도 잘 작동합니다.
- bias 옵션은 출력 shape에 영향을 주지 않습니다.

$$\begin{split} \sigma \vdash E &\Rightarrow e, c \\ e' &= e[1:k-1]@(out) \\ c' &= \{(\mathtt{rank}(e) \geq 1) \land (d_k = in)\} \\ \hline \sigma \vdash \mathtt{Linear}(in, out, bias = True)(E) \Rightarrow e', c \cup c' \end{split}$$

# Activations

torch.nn.ReLU, torch.nn.ReLU6, torch.relu, torch.nn.functional.relu





## Require

## Guarantees

• |y| = |x| (same shape)

#### Comment

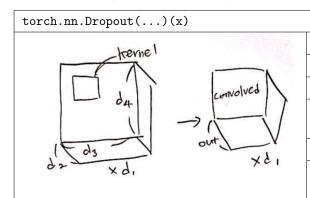
- inplace 옵션은 shape에 영향을 주지 않습니다.
- ReLU6도 ReLU와 똑같은 방식으로 shape 계산
- Bulitins인 torch.relu와 torch.nn.functional.relu는 같은거

$$\forall \mathtt{ft} \in \{\mathtt{ReLU}, \mathtt{ReLU6}\}, \quad \frac{\sigma \vdash E \Rightarrow e, c}{\sigma \vdash \mathtt{ft}(inplace = True)(E) \Rightarrow e, c}$$

$$\frac{\sigma \vdash E \Rightarrow e, c}{\sigma \vdash \mathtt{relu}(E, inplace = True) \Rightarrow e, c}$$

# Technique

torch.nn.Dropout, torch.dropout, torch.nn.functional.dropout



# Require

#### Guarantees

• |y| = |x| (same shape)

# Comment

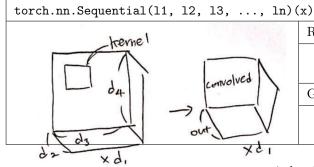
- 모든 옵션은 shape에 영향을 주지 않습니다.
- Bulitins인 torch.dropout와 torch.nn.functional.dropout는 서로 역할이 같습니다.

$$\frac{\sigma \vdash E \Rightarrow e, c}{\sigma \vdash \mathtt{Dropout}(...)(E) \Rightarrow e, c}$$

$$\frac{\sigma \vdash E \Rightarrow e, c}{\sigma \vdash \mathsf{dropout}(E, \ldots) \Rightarrow e, c}$$

# Wrapper

torch.nn.Sequential



# Require

• 순차적으로 shape이 맞아떨어져야함

## Guarantees

•  $|y| = |l_n \circ l_{n-1} \circ \cdots \circ l_1(x)|$ 

$$\frac{\sigma \vdash l_n \circ l_{n-1} \circ \cdots l_1(E) \Rightarrow e, c}{\sigma \vdash \mathtt{Sequential}(l_1, l_2, \dots, l_n)(E) \Rightarrow e, c}$$