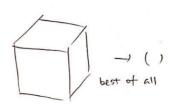
# **Statistics**

torch.max, torch.min

## torch.max(input) or .min



## Require

#### Guarantees

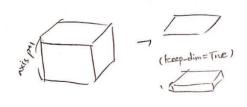
• |y| = ()

### Comment

• 주어진 텐서에서 최대/최소 값을 ()-shape 텐서로 반환(스칼라 X)

$$\forall \mathtt{ft} \in \{\mathtt{min},\mathtt{max}\}, \quad \frac{\sigma \vdash E \Rightarrow \_, c}{\sigma \vdash \mathtt{ft}(E) \Rightarrow (), c}$$

torch.max(input, dim, keep\_dim=False, out=None) or .min



### Require

- $|input| = (d_1, d_2, \dots, d_k)$
- $k \ge 1, \ 0 \le dim < k$

### Guarantees

- $|y| = |z| = (d_1, d_2, \dots, d_{dim}, d_{dim+2}, \dots, d_k)$ 인 (y, z) 튜플 반환
- 세 번째 인자  $keep\_dim$ 이 True이면  $|y|=|z|=(d_1,d_2,\ldots,d_{dim},1,d_{dim+2},\ldots,d_k)$ 로 처리

#### Comment

- 주어진 텐서에서 축 상의 최대/최소값과, 그에 해당하는 인덱스 번 호를 쌍으로 엮어 튜플로 반환
- out-텐서 인자가 있는 함수

$$\begin{split} \sigma \vdash E \Rightarrow e,c \\ k = \mathtt{rank}(e) \\ e' = e[1:n]@e[n+2:k] \\ \forall \mathtt{ft} \in \{\mathtt{min},\mathtt{max}\}, \quad & \frac{c' = \{(k \geq 1) \land (0 \leq n < k)\}}{\sigma \vdash \mathtt{ft}(E,n) \Rightarrow (e',e'),c \cup c'} \end{split}$$

tuple 형태로 반환

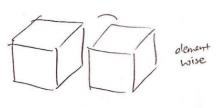
$$\begin{split} \sigma \vdash E \Rightarrow e,c \\ k = \mathtt{rank}(e) \\ e' = e[1:n]@(1)@e[n+2:k] \\ \forall \mathtt{ft} \in \{\mathtt{min},\mathtt{max}\}, \quad \frac{c' = \{(k \geq 1) \land (0 \leq n < k)\}}{\sigma \vdash \mathtt{ft}(E,n,True) \Rightarrow (e',e'),c \cup c'} \end{split}$$

tuple 형태로 반환

$$\forall \mathtt{ft} \in \{\mathtt{min},\mathtt{max}\}, \quad \frac{\sigma \vdash \mathtt{ft}(E,n) \Rightarrow (e,e), c}{\sigma \vdash \mathtt{ft}(E,n,False) \Rightarrow (e,e), c}$$

tuple 형태로 반환

torch.max(input,	other,	out=None)	or	.min



# Require

 $\bullet$  broadcastable(|input|, |other|)

### Guarantees

 $\bullet$  broadcast(|input|, |other|)

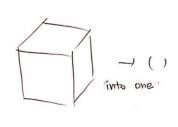
#### Comment

- 두 텐서의 elementwise 최대/최소
- out-텐서 인자가 있는 함수

$$\sigma \vdash E_1 \Rightarrow e_1, c_1$$
 
$$\sigma \vdash E_2 \Rightarrow e_2, c_2$$
 
$$\forall \texttt{ft} \in \{\texttt{min}, \texttt{max}\}, \quad \frac{\sigma \vdash E_2 \Rightarrow e_2, c_2}{\sigma \vdash \texttt{ft}(E_1, E_2) \Rightarrow broadcast(e_1, e_2), c_1 \cup c_2 \cup broadcastable(e_1, e_2)}$$

### torch.sum, torch.mean

# torch.sum(input, dtype=None) or .mean



## Require

• .mean에 대해서는 텐서 타입이 floating이어야 함

#### Guarantees

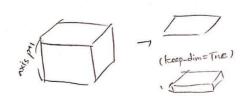
• |y| = ()

## Comment

ullet 주어진 텐서의 합계/평균값을 ()-shape 텐서로 반환(스칼라 X)

$$\forall \mathtt{ft} \in \{\mathtt{sum},\mathtt{mean}\}, \quad \frac{\sigma \vdash E \Rightarrow \_, c}{\sigma \vdash \mathtt{ft}(E) \Rightarrow (), c}$$

## torch.sum(input, dim, keep\_dim=False, dtype=None) or .mean



## Require

- $|input| = (d_1, d_2, \dots, d_k)$
- k > 1, 0 < dim < k
- .mean에 대해서는 텐서 타입이 floating이어야 함

### Guarantees

- $|y| = (d_1, d_2, \dots, d_{dim}, d_{dim+2}, \dots, d_k)$
- 세 번째 인자  $keep\_dim$ 이 True이면  $|y|=(d_1,d_2,\ldots,d_{dim},1,d_{dim+2},\ldots,d_k)$

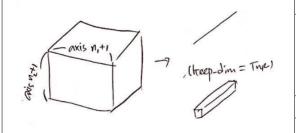
### Comment

• 주어진 텐서에서 축 상의 합/평균을 반환

$$\begin{split} \sigma \vdash E \Rightarrow e, c \\ k = \mathrm{rank}(e) \\ e' = e[1:n]@e[n+2:k] \\ \hline \sigma \vdash \mathrm{ft}(E,n) \Rightarrow e', c \cup c' \end{split}$$

$$\begin{split} \sigma \vdash E \Rightarrow e,c \\ k = \mathtt{rank}(e) \\ e' = e[1:n]@(1)@e[n+2:k] \\ \forall \mathtt{ft} \in \{\mathtt{sum},\mathtt{mean}\}, \quad \frac{c' = \{(k \geq 1) \land (0 \leq n < k)\}}{\sigma \vdash \mathtt{ft}(E,n,True) \Rightarrow e',c \cup c'} \end{split}$$

### torch.sum(input, [n1, n2, ..., n1], keep\_dim=False, dtype=None) or .mean



### Require

- $|x| = (d_1, d_2, \dots, d_k)$
- $k \ge 1, \ 0 \le n_1, n_2, \dots, n_l < k$
- .mean에 대해서는 텐서 타입이 floating이어야 함

#### Guarantees

- $|y|=(d_{i_1},d_{i_2},\dots,d_{i_{k-l}})$   $-1,2,\dots,k$ 에서  $n_1+1,n_2+1,\dots,n_l+1$ 번째 항이 지워진 shape
- 세 번째 인자 keep\_dim이 True이면 n<sub>1</sub> + 1, n<sub>2</sub> + 1,..., n<sub>l</sub> + 1번째 항은 삭제되지 않고 1로 남음

#### Comment

- 주어진 텐서에서 여러 축을 통합한 합/평균을 반환
- $[n_1, n_2, \ldots, n_l]$ 에 중복된 원소가 들어가도 상관없이 잘 작동함

$$\sigma \vdash E \Rightarrow e, c \\ k = \operatorname{rank}(e) \\ e_1 = \operatorname{if} \ 0 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[1]) \\ e_2 = \operatorname{if} \ 1 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[2]) \\ \dots \\ e_k = \operatorname{if} \ k - 1 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@ \cdots @e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \dots, r, \ 0 \leq n_i < k)\} \\ \hline \sigma \vdash \operatorname{ft}(E, (n_1, n_2, \dots, n_r)) \Rightarrow e', c \cup c' \\ \\ \sigma \vdash E \Rightarrow e, c \\ k = \operatorname{rank}(e) \\ e_1 = \operatorname{if} \ 0 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[1]) \\ e_2 = \operatorname{if} \ 1 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[2])$$

 $e_k = \text{if } k - 1 \in \{n_1, n_2, \dots, n_r\} \text{ then } (1) \text{ else } (e[k])$ 

$$\forall \mathtt{ft} \in \{\mathtt{sum}, \mathtt{mean}\}, \qquad \frac{c' = \{(k \geq 1) \land (\forall i = 1, 2, \dots, r, \ 0 \leq n_i < k)\}}{\sigma \vdash \mathtt{ft}(E, (n_1, n_2, \dots, n_r), True) \Rightarrow e', c \cup c'}$$

 $e' = e_1@e_2@\cdots@e_k$ 

## torch.prod

torch.prod(input, dtype=None),
torch.prod(input, dim, keepdim=False, dtype=None)

#### Comment

- 대부분 sum과 똑같음
- 하지만, dim 부분에 튜플값이 들어갈 수 없음. 즉 여러 축에 대한 곱을 계산할 수 없음

$$\frac{\sigma \vdash E \Rightarrow \_, c}{\sigma \vdash \operatorname{prod}(E) \Rightarrow (), c}$$

$$\begin{split} \sigma &\vdash E \Rightarrow e, c \\ k &= \mathtt{rank}(e) \\ e' &= e[1:n]@e[n+2:k] \\ c' &= \{(k \geq 1) \land (0 \leq n < k)\} \\ \hline \sigma &\vdash \mathtt{prod}(E, n) \Rightarrow e', c \cup c' \end{split}$$

$$\begin{split} \sigma &\vdash E \Rightarrow e, c \\ k &= \mathtt{rank}(e) \\ e' &= e[1:n]@(1)@e[n+2:k] \\ \underline{c'} &= \{(k \geq 1) \land (0 \leq n < k)\} \\ \overline{\sigma} &\vdash \mathtt{prod}(E, n, True) \Rightarrow e', c \cup c' \end{split}$$

### torch.norm

torch.norm(input, p='fro', dim=None, keepdim=False, out=None, dtype=None)

### Require

- $|input| = (d_1, d_2, \dots, d_k)$
- dim = None or  $0 \le dim < k$  or  $dim = (p_1, p_2, \dots, p_m)$ such that  $0 \le p_1, p_2, \dots, p_m < k$  (tuple)

#### Guarantees

• If dim = None then

$$-|y|=()$$

- If  $dim \neq None$  then
  - If keepdim = False then

$$* |y| = (d_1, d_2, \dots, d_{dim}, d_{dim+2}, \dots, d_k)$$

- If keepdim = True then

$$* |y| = (d_1, d_2, \dots, d_{dim}, 1, d_{dim+2}, \dots, d_k)$$

• 만약 *dim*이 여러 axis로 구성된 튜플이면 해당 축들이 지워지거나, 아니면 1차원으로 바뀐 텐서 shape을 반환

### Comment

- 일단 기본적인 shape 연산은 sum과 비슷함
- out-텐서 인자가 있는 함수

$$\frac{\sigma \vdash E \Rightarrow \_, c}{\sigma \vdash \mathsf{norm}(E) \Rightarrow (), c}$$

$$\sigma \vdash E \Rightarrow e, c \\ k = \operatorname{rank}(e) \\ e' = e[1:n]@e[n + 2:k] \\ c' = \{(k \geq 1) \land (0 \leq n < k)\} \\ \hline \sigma \vdash \operatorname{norm}(E, p = `fro', n, False) \Rightarrow e', c \cup c' \\ \\ \sigma \vdash E \Rightarrow e, c \\ k = \operatorname{rank}(e) \\ e' = e[1:n]@(1)@e[n + 2:k] \\ c' = \{(k \geq 1) \land (0 \leq n < k)\} \\ \hline \sigma \vdash \operatorname{norm}(E, p = `fro', n, True, \ldots) \Rightarrow e', c \cup c' \\ \\ \sigma \vdash E \Rightarrow e, c \\ k = \operatorname{rank}(e) \\ e_1 = \operatorname{if} \ 0 \in \{n_1, n_2, \ldots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[1]) \\ e_2 = \operatorname{if} \ 1 \in \{n_1, n_2, \ldots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[2]) \\ \ldots \\ e_k = \operatorname{if} \ k - 1 \in \{n_1, n_2, \ldots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n), False, \ldots) \Rightarrow e', c \cup c' \\ \\ \sigma \vdash \operatorname{norm}(E, p = `fro', (n_1, n_2, \ldots, n_r), \operatorname{then} \ (1) \ \operatorname{else} \ (e[1]) \\ e_2 = \operatorname{if} \ 1 \in \{n_1, n_2, \ldots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[2]) \\ \ldots \\ e_k = \operatorname{if} \ k - 1 \in \{n_1, n_2, \ldots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ c' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ e' = \{(k \geq 1) \land (\forall i = 1, 2, \ldots, r_n\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@\cdots@e_k \\ e' = e'1@e_2@\cdots@e_k \\ e' = e'1@e_$$

torch. Tensor.all, torch. Tensor.any

a.all(), a.all(dim, keepdim=False, out=None) or .any

Comment

shape의 기능은 prod과 똑같음
torch.Tensor 하위 함수이며, bool-타입 텐서에서만 사용 가능

$$\forall \mathtt{ft} \in \{\mathtt{all}, \mathtt{any}\}, \quad \frac{\sigma \vdash E \Rightarrow \_, c}{\sigma \vdash E.\mathtt{ft}() \Rightarrow (), c}$$

$$\begin{split} \sigma \vdash E \Rightarrow e,c \\ k = \mathtt{rank}(e) \\ e' = e[1:n]@e[n+2:k] \\ \forall \mathtt{ft} \in \{\mathtt{all},\mathtt{any}\}, \quad & \frac{c' = \{(k \geq 1) \land (0 \leq n < k)\}}{\sigma \vdash E.\mathtt{ft}(n) \Rightarrow e',c \cup c'} \end{split}$$

$$\begin{split} \sigma \vdash E \Rightarrow e,c \\ k = \mathtt{rank}(e) \\ e' = e[1:n]@(1)@e[n+2:k] \\ \forall \mathtt{ft} \in \{\mathtt{all},\mathtt{any}\}, \quad & \frac{c' = \{(k \geq 1) \land (0 \leq n < k)\}}{\sigma \vdash E.\mathtt{ft}(n,True) \Rightarrow e',c \cup c'} \end{split}$$

## torch.var, torch.std

| Require | Require | • 텐서 타입이 floating이어야 함 | Guarantees | • |y| = () | Comment | • 주어진 텐서의 분산/표준편차를 ()-shape 텐서로 반환(스칼라 X)

$$\forall \mathtt{ft} \in \{\mathtt{var},\mathtt{std}\}, \quad \frac{\sigma \vdash E \Rightarrow \_, c}{\sigma \vdash \mathtt{ft}(E) \Rightarrow (), c}$$

torch.var(input, dim, keepdim=False, unbiased=None, out=None) or torch.std(input, dim, unbiased=True, keepdim=False, out=None)

## Require

- $|input| = (d_1, d_2, \dots, d_k)$
- $k \ge 1, \ 0 \le dim < k$
- 텐서 타입이 floating이어야 함

#### Guarantees

- $|y| = (d_1, d_2, \dots, d_{dim}, d_{dim+2}, \dots, d_k)$
- $keepdim \circ | True \circ | \mathfrak{B} | | y | = (d_1, d_2, \dots, d_{dim}, 1, d_{dim+2}, \dots, d_k)$

## Comment

- 주어진 텐서에서 축 상의 분산/표준편차를 반환
- sum, mean과 마찬가지로 dim 부분은 튜플로 쓰일 수도 있음 (여러 축에 대한 분산/표춘편차)

$$\begin{split} \sigma \vdash E \Rightarrow e,c \\ k = \operatorname{rank}(e) \\ e' = e[1:n]@e[n+2:k] \\ \hline c' = \{(k \geq 1) \land (0 \leq n < k)\} \\ \hline \sigma \vdash \operatorname{ft}(E,n) \Rightarrow e',c \cup c' \\ \\ \sigma \vdash E \Rightarrow e,c \\ k = \operatorname{rank}(e) \\ e' = e[1:n]@(1)@e[n+2:k] \\ \hline c' = \{(k \geq 1) \land (0 \leq n < k)\} \\ \hline \sigma \vdash \operatorname{ft}(E,n,True) \Rightarrow e',c \cup c' \\ \end{split}$$

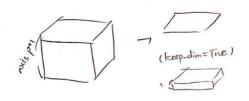
$$k = \operatorname{rank}(e) \\ e_1 = \operatorname{if} \ 0 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[1]) \\ e_2 = \operatorname{if} \ 1 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[2]) \\ \dots \\ e_k = \operatorname{if} \ k - 1 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ () \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@ \cdots @e_k \\ c' = \{(k \geq 1) \wedge (\forall i = 1, 2, \dots, r, \ 0 \leq n_i < k)\} \\ \hline \sigma \vdash \operatorname{ft}(E, (n_1, n_2, \dots, n_r)) \Rightarrow e', c \cup c' \\ \\ \sigma \vdash E \Rightarrow e, c \\ k = \operatorname{rank}(e) \\ e_1 = \operatorname{if} \ 0 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[1]) \\ e_2 = \operatorname{if} \ 1 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[2]) \\ \dots \\ e_k = \operatorname{if} \ k - 1 \in \{n_1, n_2, \dots, n_r\} \ \operatorname{then} \ (1) \ \operatorname{else} \ (e[k]) \\ e' = e_1@e_2@ \cdots @e_k \\ c' = \{(k \geq 1) \wedge (\forall i = 1, 2, \dots, r, \ 0 \leq n_i < k)\} \\ \hline \sigma \vdash \operatorname{ft}(E, (n_1, n_2, \dots, n_r), True) \Rightarrow e', c \cup c'$$

 $\sigma \vdash E \Rightarrow e, c$ 

keepdim이 True인 상황에 대한 proof tree

## torch.mode, torch.median

## torch.mode(input, dim=-1, keep\_dim=False, out=None) or torch.median



#### Require

- $|input| = (d_1, d_2, \dots, d_k)$
- *k* > 1
- $0 \le dim < k \ (dim \leftarrow k 1 \ \text{if} \ dim = -1)$

# Guarantees

- $|y| = |z| = (d_1, d_2, \dots, d_{dim}, d_{dim+2}, \dots, d_k)$ 인 (y, z) 튜플 반환
- 세 번째 인자  $keep\_dim$ 이 True이면  $|y|=|z|=(d_1,d_2,\ldots,d_{dim},1,d_{dim+2},\ldots,d_k)$

#### Comment

- 입력받은 축을 기준으로한 통계적 최빈값/중간값 계산 함수
- (결과값, 인덱스) 튜플 형태로 반환하기 때문에, out인자도 들어가는 텐서도 2-원소 튜플 형태로 들어감

$$\begin{split} \sigma \vdash E \Rightarrow e, c \\ k = \texttt{rank}(e) \\ e' = e[1:n]@e[n+2:k] \\ \forall \texttt{ft} \in \{\texttt{mode}, \texttt{median}\}, \quad \frac{c' = \{(k \geq 1) \land (0 \leq n < k)\}}{\sigma \vdash \texttt{ft}(E, n = -1, out = None) \Rightarrow (e', e'), c \cup c'} \end{split}$$

tuple 형태로 반환

$$\begin{split} \sigma \vdash E \Rightarrow e, c \\ k = \mathtt{rank}(e) \\ e' = e[1:n]@(1)@e[n+2:k] \\ \forall \mathtt{ft} \in \{\mathtt{mode}, \mathtt{median}\}, \quad \frac{c' = \{(k \geq 1) \land (0 \leq n < k)\}}{\sigma \vdash \mathtt{ft}(E, n = -1, True, out = None) \Rightarrow (e', e'), c \cup c'} \end{split}$$

tuple 형태로 반환