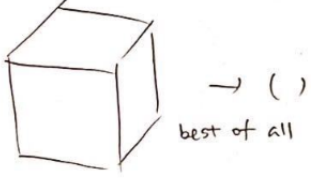
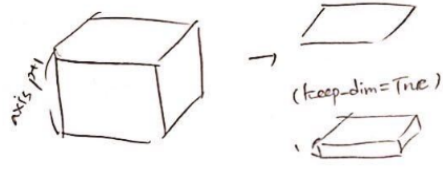


Statistics

torch.max, torch.min

| torch.max(input) or .min | |
|---|--|
|  | Require |
| | |
| | Guarantees |
| | <ul style="list-style-type: none"> • $y = ()$ |
| | Comment |
| | <ul style="list-style-type: none"> • 주어진 텐서에서 최대/최소 값을 ()-shape 텐서로 반환(스칼라 X) |

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

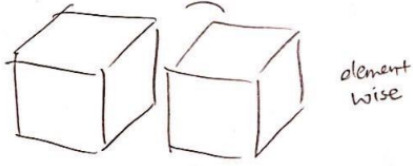
| torch.max(input, dim, keep_dim=False, out=None) or .min | |
|--|---|
|  | Require |
| | <ul style="list-style-type: none"> • $input = (d_1, d_2, \dots, d_k)$ • $k \geq 1, 0 \leq \text{dim} < k$ |
| | Guarantees |
| | <ul style="list-style-type: none"> • $y = z = (d_1, d_2, \dots, d_{\text{dim}}, d_{\text{dim}+2}, \dots, d_k)$인 (y, z) 튜플 반환 • 세 번째 인자 <code>keep_dim</code>이 <code>True</code>이면 $y = z = (d_1, d_2, \dots, d_{\text{dim}}, 1, d_{\text{dim}+2}, \dots, d_k)$로 처리 |
| | Comment |
| | <ul style="list-style-type: none"> • 주어진 텐서에서 축 상의 최대/최소값과, 그에 해당하는 인덱스 번호를 쌍으로 묶어 튜플로 반환 • <code>out</code>-텐서 인자가 있는 함수 |

$$\forall \text{ft} \in \{\min, \max\}, \quad \frac{\begin{array}{l} \sigma \vdash E \Rightarrow e, c \\ k = \text{rank}(e) \\ e' = e[1:n] @ e[n+2:k] \\ c' = \{(k \geq 1) \wedge (0 \leq n < k)\} \end{array}}{\sigma \vdash \text{ft}(E, n) \Rightarrow (e', e'), c \cup c'} \quad \text{tuple 형태로 반환}$$

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

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

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



Require

- $\text{broadcastable}(|input|, |other|)$

Guarantees

- $\text{broadcast}(|input|, |other|)$

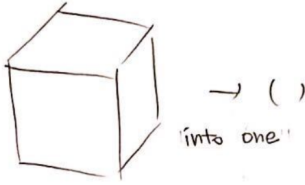
Comment

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

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

torch.sum, torch.mean

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



Require

- *.mean*에 대해서는 텐서 타입이 floating이어야 함

Guarantees

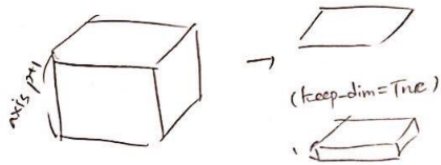
- $|y| = ()$

Comment

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

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

torch.sum(input, dim, keep_dim=False, dtype=None) or .mean



Require

- $|input| = (d_1, d_2, \dots, d_k)$
- $k \geq 1, 0 \leq \text{dim} < k$
- *.mean*에 대해서는 텐서 타입이 floating이어야 함

Guarantees

- $|y| = (d_1, d_2, \dots, d_{\text{dim}}, d_{\text{dim}+2}, \dots, d_k)$
- 세 번째 인자 *keep_dim*이 *True*이면 $|y| = (d_1, d_2, \dots, d_{\text{dim}}, 1, d_{\text{dim}+2}, \dots, d_k)$

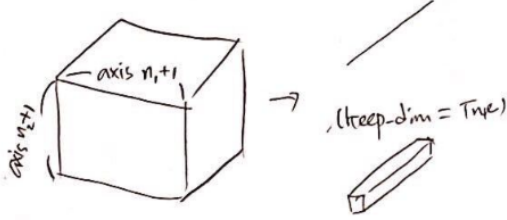
Comment

- 주어진 텐서에서 축 상의 합/평균을 반환

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

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

`torch.sum(input, [n1, n2, ..., nl], keep_dim=False, dtype=None) or .mean`



Require

- $|x| = (d_1, d_2, \dots, d_k)$
- $k \geq 1, 0 \leq n_1, n_2, \dots, n_l < k$
- `.mean`에 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_dim`이 `True`이면 $n_1+1, n_2+1, \dots, n_l+1$ 번째 항은 삭제되지 않고 1로 남음

Comment

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

$$\begin{array}{l}
\sigma \vdash E \Rightarrow e, c \\
k = \text{rank}(e) \\
e_1 = \text{if } 0 \in \{n_1, n_2, \dots, n_r\} \text{ then } () \text{ else } (e[1]) \\
e_2 = \text{if } 1 \in \{n_1, n_2, \dots, n_r\} \text{ then } () \text{ else } (e[2]) \\
\vdots \\
e_k = \text{if } k-1 \in \{n_1, n_2, \dots, n_r\} \text{ then } () \text{ else } (e[k]) \\
e' = e_1 @ e_2 @ \dots @ e_k \\
c' = \{(k \geq 1) \wedge (\forall i = 1, 2, \dots, r, 0 \leq n_i < k)\} \\
\forall \text{ft} \in \{\text{sum}, \text{mean}\}, \quad \frac{}{\sigma \vdash \text{ft}(E, (n_1, n_2, \dots, n_r)) \Rightarrow e', c \cup c'}
\end{array}$$

$$\begin{array}{l}
\sigma \vdash E \Rightarrow e, c \\
k = \text{rank}(e) \\
e_1 = \text{if } 0 \in \{n_1, n_2, \dots, n_r\} \text{ then } (1) \text{ else } (e[1]) \\
e_2 = \text{if } 1 \in \{n_1, n_2, \dots, n_r\} \text{ then } (1) \text{ else } (e[2]) \\
\vdots \\
e_k = \text{if } k-1 \in \{n_1, n_2, \dots, n_r\} \text{ then } (1) \text{ else } (e[k]) \\
e' = e_1 @ e_2 @ \dots @ e_k \\
c' = \{(k \geq 1) \wedge (\forall i = 1, 2, \dots, r, 0 \leq n_i < k)\} \\
\forall \text{ft} \in \{\text{sum}, \text{mean}\}, \quad \frac{}{\sigma \vdash \text{ft}(E, (n_1, n_2, \dots, n_r), \text{True}) \Rightarrow e', c \cup c'}
\end{array}$$

torch.prod

| | |
|---|--|
| <code>torch.prod(input, dtype=None),</code> <code>torch.prod(input, dim, keepdim=False, dtype=None)</code> | |
| | Comment |
| | <ul style="list-style-type: none"> 대부분 <code>sum</code>과 똑같음 하지만, <code>dim</code> 부분에 튜플값이 들어갈 수 없음. 즉 여러 축에 대한 곱을 계산할 수 없음 |

$$\frac{\sigma \vdash E \Rightarrow \neg, c}{\sigma \vdash \text{prod}(E) \Rightarrow (), c}$$

$$\frac{\begin{array}{l} \sigma \vdash E \Rightarrow e, c \\ k = \text{rank}(e) \\ e' = e[1:n]@e[n+2:k] \\ c' = \{(k \geq 1) \wedge (0 \leq n < k)\} \end{array}}{\sigma \vdash \text{prod}(E, n) \Rightarrow e', c \cup c'}$$

$$\frac{\begin{array}{l} \sigma \vdash E \Rightarrow e, c \\ k = \text{rank}(e) \\ e' = e[1:n]@(1)@e[n+2:k] \\ c' = \{(k \geq 1) \wedge (0 \leq n < k)\} \end{array}}{\sigma \vdash \text{prod}(E, n, \text{True}) \Rightarrow e', c \cup c'}$$

torch.Tensor.all, torch.Tensor.any

| | |
|--|--|
| <code>a.all()</code> , <code>a.all(dim, keepdim=False, out=None)</code> or <code>.any</code> | |
| | Comment |
| | <ul style="list-style-type: none"> <code>shape</code>의 기능은 <code>prod</code>과 똑같음 <code>torch.Tensor</code> 하위 함수이며, bool-타입 텐서에서만 사용 가능 |

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

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

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

torch.var, torch.std

| torch.var(input, unbiased=True) or .std | |
|---|---|
| | Require |
| | <ul style="list-style-type: none"> • 텐서 타입이 floating이어야 함 |
| | Guarantees |
| | <ul style="list-style-type: none"> • $y = ()$ |
| | Comment |
| | <ul style="list-style-type: none"> • 주어진 텐서의 분산/표준편차를 ()-shape 텐서로 반환(스칼라 X) |

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

| torch.var(input, dim, keepdim=False, unbiased=None, out=None) or torch.std(input, dim, unbiased=True, keepdim=False, out=None) | |
|---|---|
| | Require |
| | <ul style="list-style-type: none"> • $input = (d_1, d_2, \dots, d_k)$ • $k \geq 1, 0 \leq dim < k$ • 텐서 타입이 floating이어야 함 |
| | Guarantees |
| | <ul style="list-style-type: none"> • $y = (d_1, d_2, \dots, d_{dim}, d_{dim+2}, \dots, d_k)$ • $keepdim$이 <i>True</i>이면 $y = (d_1, d_2, \dots, d_{dim}, 1, d_{dim+2}, \dots, d_k)$ |
| | Comment |
| | <ul style="list-style-type: none"> • 주어진 텐서에서 축 상의 분산/표준편차를 반환 • <code>sum</code>, <code>mean</code>과 마찬가지로 dim 부분은 튜플로 쓰일 수도 있음 (여러 축에 대한 분산/표준편차) |

$$\forall \text{ft} \in \{\text{var}, \text{std}\}, \quad \frac{\begin{array}{l} \sigma \vdash E \Rightarrow e, c \\ k = \text{rank}(e) \\ e' = e[1:n] @ e[n+2:k] \\ c' = \{(k \geq 1) \wedge (0 \leq n < k)\} \end{array}}{\sigma \vdash \text{ft}(E, n) \Rightarrow e', c \cup c'}$$

$$\forall \text{ft} \in \{\text{var}, \text{std}\}, \quad \frac{\begin{array}{l} \sigma \vdash E \Rightarrow e, c \\ k = \text{rank}(e) \\ e' = e[1:n] @ (1) @ e[n+2:k] \\ c' = \{(k \geq 1) \wedge (0 \leq n < k)\} \end{array}}{\sigma \vdash \text{ft}(E, n, \text{True}) \Rightarrow e', c \cup c'}$$

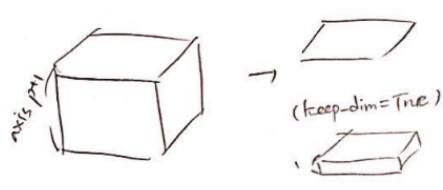
$keepdim$ 이 *True*인 상황에 대한 proof tree

$$\begin{array}{l}
\sigma \vdash E \Rightarrow e, c \\
k = \text{rank}(e) \\
e_1 = \text{if } 0 \in \{n_1, n_2, \dots, n_r\} \text{ then } () \text{ else } (e[1]) \\
e_2 = \text{if } 1 \in \{n_1, n_2, \dots, n_r\} \text{ then } () \text{ else } (e[2]) \\
\vdots \\
e_k = \text{if } k-1 \in \{n_1, n_2, \dots, n_r\} \text{ then } () \text{ else } (e[k]) \\
e' = e_1 @ e_2 @ \dots @ e_k \\
c' = \{(k \geq 1) \wedge (\forall i = 1, 2, \dots, r, 0 \leq n_i < k)\} \\
\hline
\forall \text{ft} \in \{\text{var}, \text{std}\}, \quad \sigma \vdash \text{ft}(E, (n_1, n_2, \dots, n_r)) \Rightarrow e', c \cup c'
\end{array}$$

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

*keepdim*이 *True*인 상황에 대한 proof tree

torch.mode, torch.median

| torch.mode(input, dim=-1, keep_dim=False, out=None) or torch.median | |
|---|---|
|  | Require |
| | <ul style="list-style-type: none"> • $input = (d_1, d_2, \dots, d_k)$ • $k \geq 1$ • $0 \leq dim < k$ ($dim \leftarrow k-1$ if $dim = -1$) |
| | Guarantees |
| | <ul style="list-style-type: none"> • $y = z = (d_1, d_2, \dots, d_{dim}, d_{dim+2}, \dots, d_k)$인 (y, z) 튜플 반환 • 세 번째 인자 <i>keep_dim</i>이 <i>True</i>이면 $y = z = (d_1, d_2, \dots, d_{dim}, 1, d_{dim+2}, \dots, d_k)$ |
| | Comment |
| | <ul style="list-style-type: none"> • 입력받은 축을 기준으로한 통계적 최빈값/중간값 계산 함수 • (결과값, 인덱스) 튜플 형태로 반환하기 때문에, out인자도 들어가는 텐서도 2-원소 튜플 형태로 들어감 |

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

tuple 형태로 반환

$$\begin{array}{c}
\sigma \vdash E \Rightarrow e, c \\
k = \mathbf{rank}(e) \\
e' = e[1:n]@ (1) @ e[n+2:k] \\
c' = \{(k \geq 1) \wedge (0 \leq n < k)\} \\
\forall \mathbf{ft} \in \{\mathbf{mode}, \mathbf{median}\}, \quad \frac{}{\sigma \vdash \mathbf{ft}(E, n = -1, \mathit{True}, \mathit{out} = \mathit{None}) \Rightarrow (e', e'), c \cup c'}
\end{array}
\quad \text{tuple 형태로 반환}$$

$$\forall \mathbf{ft} \in \{\mathbf{mode}, \mathbf{median}\}, \quad \frac{\sigma \vdash \mathbf{ft}(E, n) \Rightarrow (e, e), c}{\sigma \vdash \mathbf{ft}(E, n = -1, \mathit{False}, \mathit{out} = \mathit{None}) \Rightarrow (e, e), c}
\quad \text{tuple 형태로 반환}$$