線性代數沒有明確定義,但重要的Python張量操作!!

import numpy as np
import tensorflow as tf
import torch as tc

1.張量 X 資料結構間互相轉換

X>	list	numpy	pytorch	tensorflow
list	-	np.array(X)	tc.tensor(X)	tf.convert_to_tensor(X)
numpy	X.tolist()	-	tc.tensor(X)	tf.convert_to_tensor(X)
pytorch	X.tolist()	X.numpy()	-	-
tensorflow	-	X.numpy()	-	-

2.常見張量形狀操作!!

A, B表示 tensor 名稱

名稱	numpy	torch	tensorflow
取得張量形狀	A.shape	A.shape / A.size()	A.shape
取得張量大小	A.size	A.numel()	tf.size(A)
轉置	np.transpose(A,[1,0])	A.permute(1,0)	tf.transpose(A, perm=[1,0])
攤平至1維	A.flatten()	tc.flatten(A)	-
重訂形狀	A.reshape(m,n)	A.view(m,n)	tf.reshape(A,[m,n])
合併張量	np.concatenate([A,B],axis=1)	tc.cat([A,B],dim=1)	tf.concat([A, B], axis=1)
堆積張量	np.stack([A,B],axis=1)	tc.stack([A,B],dim=1)	tf.stack([A,B], axis=1)
壓縮維度 (去1)	np.squeeze(A,axis=0)	tc.squeeze(A,dim=0)	tf.squeeze(A,axis=1)
提升維度 (加1)	np.unsqueeze(A,axis=0)	tc.unsqueeze(A,dim=0)	tf.unsqueeze(A,axis=1)
內存連續化	np.ascontiguousarray(A)	A.contiguous()	-

轉置:

\$\$

 $A^{T}:= transpose \left(A \cdot (A_{ijk}), [2,0,1] \cdot (A_{kij}) \right)$

 $A_{.shape} = (I,J,K) \quad A^{T}_{.shape} = (K,I,J) \$

=======備註: \color{red}{需再使用"內存連續化"才能讓空間儲存連續} ========= \\

\$\$

$$egin{aligned} A^T := transpose igg(A \equiv (A_{ijk}), [2,0,1]igg) &= (A_{kij}) \ A_{.shape} = (I,J,K) \quad A_{.shape}^T = (K,I,) \end{aligned}$$

$$egin{aligned} A^T := transposeigg(A \equiv (A_{ijk}), [2,0,1]igg) = (A_{kij}) \ A_{.shape} = (I,J,K) \quad A_{.shape}^T = (K,I,J) \end{aligned}$$

攤平至1維:

$$B := flatten(A) \Longrightarrow B_{(iJK+jK+k)} := A_{ijk}$$

重訂形狀:

$$\begin{split} D := reshape(A, [I', J', K', T']) &\Longrightarrow D_{i'j'k't'} = C_{(i'J'K'T'+j'K'T'+k'T'+t')} = B_{(iJK+jK+k)} = A_{ijk} \\ &= = = = = = = \# \ \, \sharp : \ \, \text{fold maps} \, \text{fold map$$

合併張量:

$$A_{.shape} = (I, J_1, K) \quad B_{.shape} = (I, J_2, K)$$
 $C := concat([A, B], dim = 1)$
 $C_{.shape} = (I, J_1 + J_2, K)$

堆積張量:

$$egin{aligned} A_{.shape} &= (I,J,K) & B_{.shape} &= (I,J,K) \ C &:= stack([A,B],dim=1) \ D &:= stack([A,B,A],dim=0) \ E &:= stack([A,A,A,A],dim=3) \ C_{.shape} &= (I,2,J,K) \ D_{.shape} &= (3,I,J,K) \ E_{.shape} &= (I,J,K,4) \end{aligned}$$

壓縮維度:

$$egin{aligned} A_{.shape} &= (extbf{1}, I, J, extbf{1}, K) \ B &:= squeeze(A, dim = 0) \ C &:= squeeze(A, dim = [0, 3]) \ B_{.shape} &= (I, J, extbf{1}, K) \ C_{.shape} &= (I, J, K) \end{aligned}$$

提升維度:

$$egin{aligned} A_{.shape} &= (I,J,K) \ B := unsqueeze(A,dim=0) \ C := unsqueeze(A,dim=2) \ B_{.shape} &= (extbf{1},I,J,K) \ C_{.shape} &= (I,J, extbf{1},K) \end{aligned}$$

3. BroadCasting (定義不同形狀的張量如何運算!!)

- Numpy: https://docs.scipy.org/doc/numpy/user/basics.broadcasting.html
- Pytorch: https://pytorch.org/docs/stable/notes/broadcasting.html
- Tensorflow: https://www.tensorflow.org/xla/broadcasting

$$A_{.shape} = (8, 3, 4, 1, 5), B_{.shape} = (3, 1, 4, 5)$$

 $C := A \oplus B \Longrightarrow C_{.shape} = (8, 3, 4, 4, 5)$