

Deep Learning Frameworks

~~TOPIC~~

In Deep learning we have two main deep learning frameworks.

1 → TensorFlow

2 → Pytorch

① TensorFlow

→ it is an open source machine learning library for preprocessing data, modeling data and training models

↳ it contains many of most common machine learning functions you will want to use.

introduction to tensors

→ if you have ever used numpy, tensors are like ~~no~~ numpy arrays.

→ for now you can think of tensors as a multi-dimensional numerical representation of somethings, this somethings can be.

- numbers

- images

- text or it could be some other form of info information you want to represent with numbers

Difference b/w tensors and numpy array is tensors can be used on GPUs and TPUs

o Creating tensors with `tf.constant()`

① $\text{Scalar} = \text{tf.constant}(8)$

↳ How constant creates a scalar value. $\text{Scalar.ndim} \rightarrow 0$ for here.

Scalar \rightarrow also known as rank 0 tensor. Because it has no dimensions (it's just a number)

② ~~Scalar~~ Vector = $\text{tf.constant}([3, 3])$

Here using constant we are creating a vector of 3 \rightarrow of 1-dim $\rightarrow (1 \times 2)$
~~of 3x3~~ \hookrightarrow creates a 2-D array consist [1, 2]

③ 2D Vector = $\text{tf.constant}([[[1, 2], [3, 4]]])$

Here we are creating a 2-D array.
by default `tf.constant()` creates core tensors with either `int32` or `float32` dtype.

But we can also specify dtype as

`dtype = tf.float16`

`tf.constant([1, 2, 3]), dtype = tf.int16`

NOTE \rightarrow We can create n-Dim of tensors and n can be any positive number

② Creating tensors with `tf.variable()`

So let's understand why we are learning `tf.variable` to create tensors which we do have `tf.constant`.

\hookrightarrow As tensors created with `tf.constant` are immutable (can't be changed), can only be used to create new tensors)

(whereas tensors created with ~~tensor~~ `tf.Variable` are mutable (can be changed))

Vector = `tf.Variable([2, 3, 4])`

Signature is same as of `Constant`.

To change value we use `.assign()`

`Vector(5).assign(11)`

③ Creating Random Vectors

we can create random tensors using
`tf.random.Generator.from_seed(42)`

④

How to shuffle the order of tensor

why we need shuffle

\hookrightarrow Let's say we are working with 15000 images first 8000 are of cat and last 7000 are of dog, if we train it on dat that way only we might get overfitting model for the orders so we shuffle it.

\rightarrow `tf.random.shuffle(vector)`

(5) other ways to create tensors

to create tensor with value 1
 $f.f.ones(\text{shape} = (3, 2))$

to create tensor with value 0
 $f.f.zeros(\text{shape} = (3, 2))$

(6) you can also convert/creates to tensors from numpy array

Only diff b/w them is tensors can be run on GPU's.

$\text{tensor} = f.f.constant(0.0, \text{shape} = (2, 4, 5))$

(element in arr should be equal to $2 \times 4 \times 3$)

(7) $f.f.transpose(tensor) \rightarrow$ gives transpose & row \Rightarrow col.

(8) $f.f.reshape(2, 4) \rightarrow$ reshapes & tensor

(9) To change D-type of tensor

$B = f.f.cast(B, \text{dtype} = f.f.int32)$

- Getting information from tensors
 - shape → length (no. of element) of each dimension of tensor
 - rank → The number of tensor dimensions
 - Axis or dimension → A particular dimension of a tensor
 - Size → The total number of items in tensor
- * Manipulating tensors with basic operation
 - ① Addition
 - tensor = tf.constant [1, 2, 3]
 - tensor + 2
 - [3, 4, 5]
 - it is giving us a copy, original remains unchanged as we have used tf.constant tensor
 - [1, 2, 3]
 - # Similarly we can do (*, -, /) etc
 - tensor * 10
 - tf.multiply(tensor, 10)
 - (10, 20, 30)
- * Manipulating with matrix multiplication
 - ① tf.multiply(tensor1, tensor2)
 - ② tensor1 @ tensor2

→ Getting absolute value

tf.abs(tensor)

get min, max, sum, mean

- $\text{tf.reduce_min(tensor)}$
- $\text{tf.reduce_max(tensor)}$
- $\text{tf.reduce_mean(tensor)}$
- $\text{tf.reduce_sum(tensor)}$
- $\text{tf.reduce_variance(tensor)}$
- $\text{tf.reduce_std(tensor)}$

to get index of min, max

tf.argmax(tensor)

tf.argmin(tensor)