



PyTorch Cheat Sheet

V2021.01.16

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What is PyTorch?

- NumPy on GPU
- Automatic differentiation library
- State of the art neural network library
- Utility libraries for vision, text and audio
- Open source machine learning framework

Torch.Tensor

- Data Type:
 - FloatTensor
 - BoolTensor
 - LongTensor
- Tensor Properties:
 - data, dtype, device, requires_grad
- Data Type Transformation:
 - float(), bool(), int()
- Numpy Array & Python List:
 - torch.tensor()
 - numpy()
 - torch.from_numpy()
 - tolist()
- Construction:
 - torch.rand(size=(3, 3))
 - torch.randn(size=(3, 3))
 - torch.arange(start=1, end=10, step=2)
 - torch.linspace(3, 10, steps=3)
 - torch.logspace(-4, 0, steps=5, base=10)
- Data Selection:
 - Same with numpy, e.g. tensor[:, 0]
- Functions:
 - max, argmax, min, argmin
 - mean, median, mode, var, quantile, sum
 - dot, mm, inverse, cholesky, svd
- Transformation:
 - view, reshape
 - unsqueeze [add dim], squeeze [remove dim if 1]
 - cat, transpose, permute

General Neural Model Steps

1. Define Neural Network Components and Network Structure

2. Set up hyperparameters, e.g. device, learning rate, epochs, etc

3. Data Loading and Training/Validation/Testing Split

4. Define loss function and Optimizer

5. Model Train

6. Plot Accuracy and Tuning

7. Model Test

8. Model Deployment

→ Code example: [basic neural network](#)

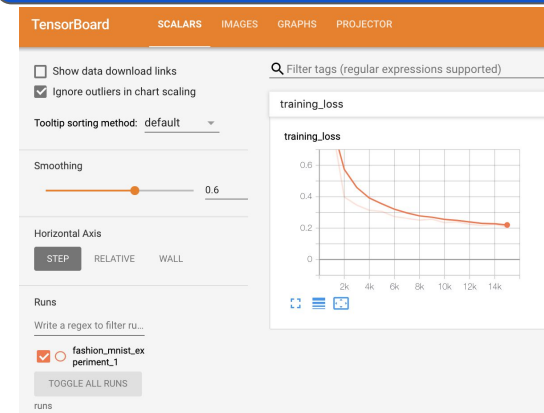
Weight Initialization

- Parameters in nn.Module have default initialization method, e.g.
 - Linear Layer: [pytorch github](#)
 - Conv Layer: [pytorch github](#)
 - BatchNorm Layer: [pytorch github](#)
- Customize Initialization
 - nn.init.uniform_(tensor, a=0.0, b=1.0)
 - nn.init.kaiming_uniform_()
 - nn.init.constant_(tensor, val)

Reproducibility

- torch.manual_seed(seed)
- numpy.random.seed(seed)
- random.seed(seed)
- torch.backends.cudnn.benchmark = False
- torch.set_deterministic(True)

Debugging: TensorBoard



tensorboard --logdir=PATH/LOG

- Plot loss time series
- Plot Images/transformed images
- Compare hyper-parameters
- Code example: [github](#)

DataSet & DataLoader

- DataSet: A class representing a dataset
 - It's common to write own DataSet
 - We could directly access DataSet by [i], but it is much common to access by Data Loader

DataSet Code Template

```
1 class MyDataSet(Dataset):
2     ...def __init__(self, csv_file, transform=None):
3     ...self.data = pd.read_csv(csv_file)
4     ...self.feature, self.target = split(self.data)
5     ...if transform:
6     ...self.feature = transform(self.feature)
7
8     ...def __len__(self):
9     ...return len(self.target)
10
11    ...def __getitem__(self, idx):
12    ...return self.feature[idx], self.target[idx]
```

Common Operations on DataSet:

- random_split()
- RandomSampler()
- Subset()

DataLoader: a smart iterator

- batch_size
- shuffle
- collate_fn

```
1 for features, target in data_loader:
2     ...pred = model(features)
3     ...cost = criterion(target, pred)
```

No need to define own Data Loader

Standard Loss Functions

Regression

- nn.MSELoss(Input, Target)
 - Input: (N) where value is any float number
 - Target: (N) where value is any float number
- nn.L1Loss(Input, Target)
 - Input: (N) where value is any float number
 - Target: (N) where value is any float number

Classification

- **Multi-class:** CrossEntropyLoss(Input, Target)
 - Input: (N,C) where C = number of classes
 - Target: (N) where each value is [0, C-1]
- **Binary-class:** BCEWithLogitsLoss(Input, Target)
 - Input: (N) where value is any float number
 - Target: (N) where each value is [0, 1]

○ *note: these two loss functions include sigmoid function already.*

Image Classification

→ torch.nn

- nn.Conv2d
 - E.g. Conv2d(in_channels=3, out_channels=8, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1))
- nn.MaxPool2d
 - E.g. MaxPool2d(kernel_size=(2, 2), stride=(2, 2))
- nn.BatchNorm2d
- nn.Sequential

→ torchvision.datasets

→ torchvision.models

- VGG
- Inception v3

→ Transfer learning: [code](#)

- model = models.vgg16() # random parameters
- model = models.vgg16(pretrained=True) # reuse initial parameters
- model.parameters().requires_grad = False # fix parameters

→ torchvision.transforms

- transforms.RandomAffine
- transforms.RandomRotation
- transforms.ToTensor
- transforms.Normalize

→ Pytorch VGG: [code](#)

Object Detection (YOLO)

- Location Encoding ([target encoding](#))
- Neural Network
 - DarkNet, RestNet, etc
- Accuracy Measurement
 - IOU ([intersection over union](#))
 - mAP ([mean average precision](#))
- Non Max Suppression ([code](#))
- YOLO Cost (classification cost & box cost)
- Image/Box Transformation
- Not much support in Pytorch yet
- Pytorch 3rd Lib: [github](#)

Neural Style Transfer

- Motivation: fix Neural Network parameters and optimize on the input image
- Cost Function:
 - Image similarity & style similarity
- Code example: [github](#)

Basic Natural Language Processing

→ torchtext.vocab.Vocab

- .freqs: the frequencies of tokens
- .stoi: mapping token strings to numerical identifiers
- .itos: numerical identifiers to strings

→ torchtext.data.Dataset

- It is different from general dataset, because it allows for each column containing sequence of features.
- It must have two functions: split() and iters()

→ torchtext.data.TabularDataset

→ torchtext.data.Field

- The concept is kind of strange, but it primarily tells whether a column is a sequence or singular value.

→ Overall, torchtext.data.dataset has a strange code flavor, an example blow:

```
# create Field objects
AUTHOR = data.Field()
AGE = data.Field()
TWEET = data.Field()
LOCATION = data.Field()

# create tuples representing the columns
fields = [
    ('author', AUTHOR),
    ('location', LOCATION),
    (None, None), # ignore age column
    ('tweet', TWEET)
]

# load the dataset in json format
train_ds, valid_ds, test_ds = data.TabularDataset.splits(
    path = 'data',
    train = 'train.tsv',
    validation = 'valid.tsv',
    test = 'test.tsv',
    format = 'tsv',
    fields = fields,
    skip_header = True
)
```

→ Tokenizer

- text.split()
- torchtext.data.get_tokenizer(text)
- [spacy](#) tokenizer

→ common NLP Datasets

- E.g. torchtext.datasets.[IMDB](#)

→ torch.nn

- nn.[RNN](#)(input_size, hidden_size, num_layers)
- nn.[GRU](#)(input_size, hidden_size, num_layers)
- nn.[LSTM](#)(input_size, hidden_size, num_layers)

→ Example NLP POC

- [bag-of-word](#) model
- [Embedding](#) (glove) model
- [GRU](#) model

Basic Audio Processing

→ torchaudio.load()

- datawaveform, sample_rate = torchaudio.load('example.mp3')

→ torchaudio.transforms

- torchaudio.transforms.MelSpectrogram (*most popular*)
 - sample_rate: int
 - win_length: int
 - hop_length: int
 - n_fft: int
 - F_min: float
 - f_max: float
- torchaudio.transforms.Spectrogram
- torchaudio.transforms.MelScale

→ torchtext.datasets

→ torchaudio.models

→ Example Audio ML POC code

- [Audio Feature Extraction](#)
- [Audio + RestNet](#)