

What you will learn



Define a data loader



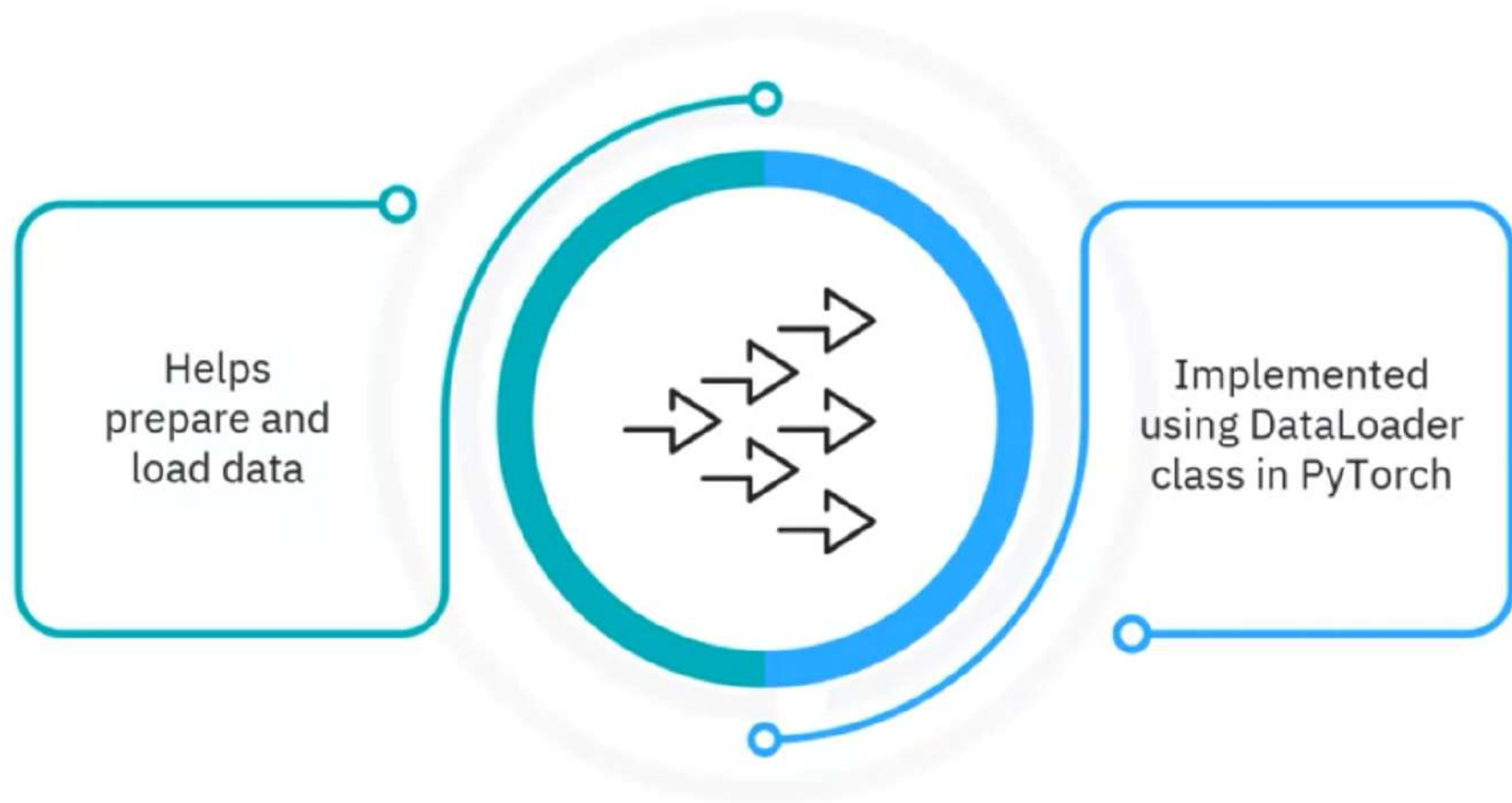
Explain its purpose



Describe the
DataLoader class and
batch functions



Data loader



Purpose of using NLP data loader

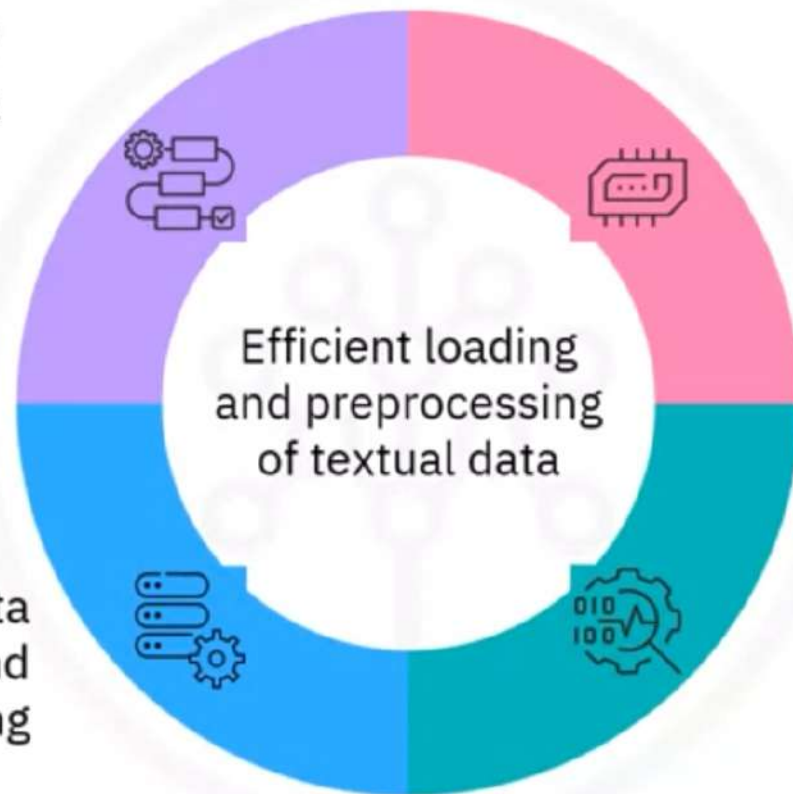
Efficient batching and shuffling of data

Memory optimization through on-the-fly preprocessing

Efficient loading and preprocessing of textual data

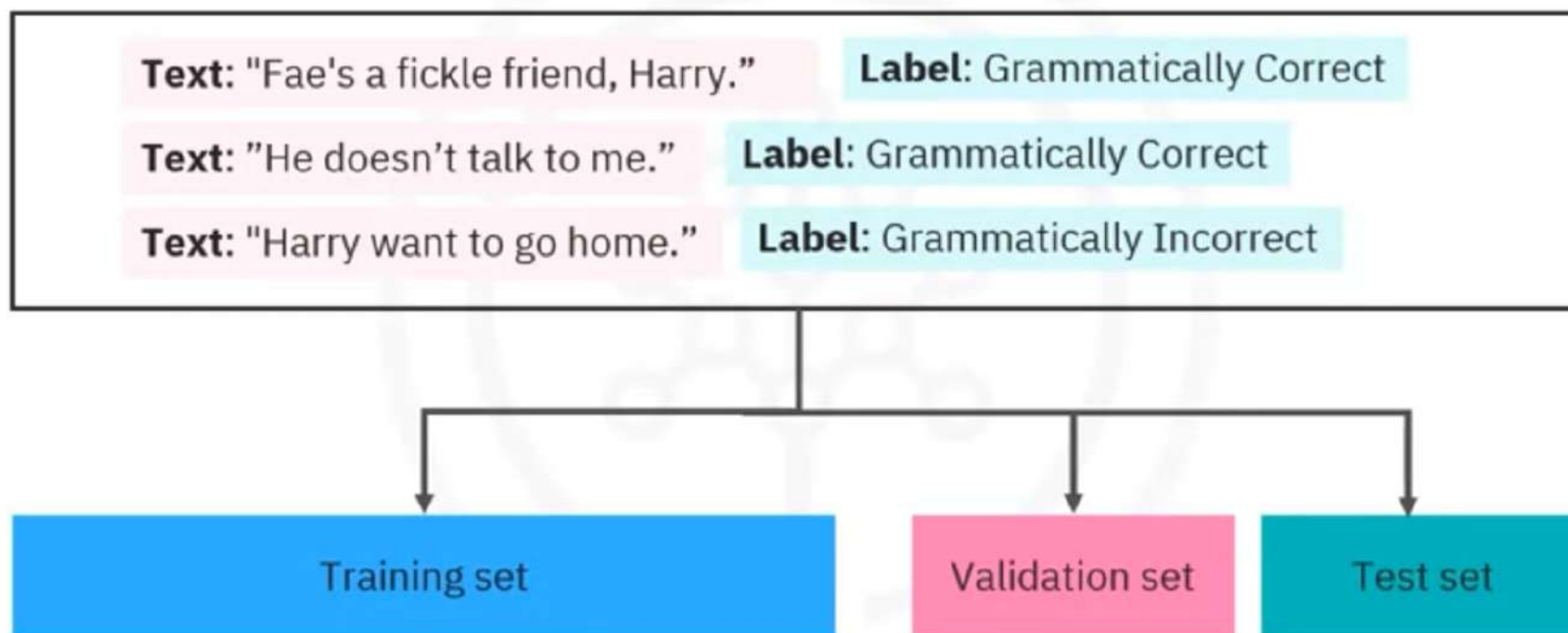
Simplified data augmentation and preprocessing

Seamless integration with PyTorch training pipeline



PyTorch data sets

- Data set: Collection of data samples and their labels



CustomDataset

```
from torch.utils.data import Dataset
```

```
sentences = [ "If you want to know what a man's like, take a good look at how he  
treats his inferiors, not his equals.", "Fae's a fickle friend, Harry.", "It is our  
choices, Harry, that show what we truly are, far more than our abilities.", "Soon we  
must all face the choice between what is right and what is easy.", "Youth cannot know  
how age thinks and feels. But old men are guilty if they forget what it was to be  
young.", "You are awesome!"]
```

```
class CustomDataset(Dataset):
```

```
    def __init__(self, sentences):  
        self.sentences = sentences
```

Downloads and reads data

```
    def __len__(self):  
        return len(self.sentences)
```

Returns the data length

```
    def __getitem__(self, idx):  
        return self.sentences[idx]
```

Returns one item on the index



CustomDataset

```
dataset=CustomDataset(sentences)
```

```
dataset[0]:
```

```
"If you want to know what a man's like, take a good look at how he treats  
his inferiors, not his equals"
```

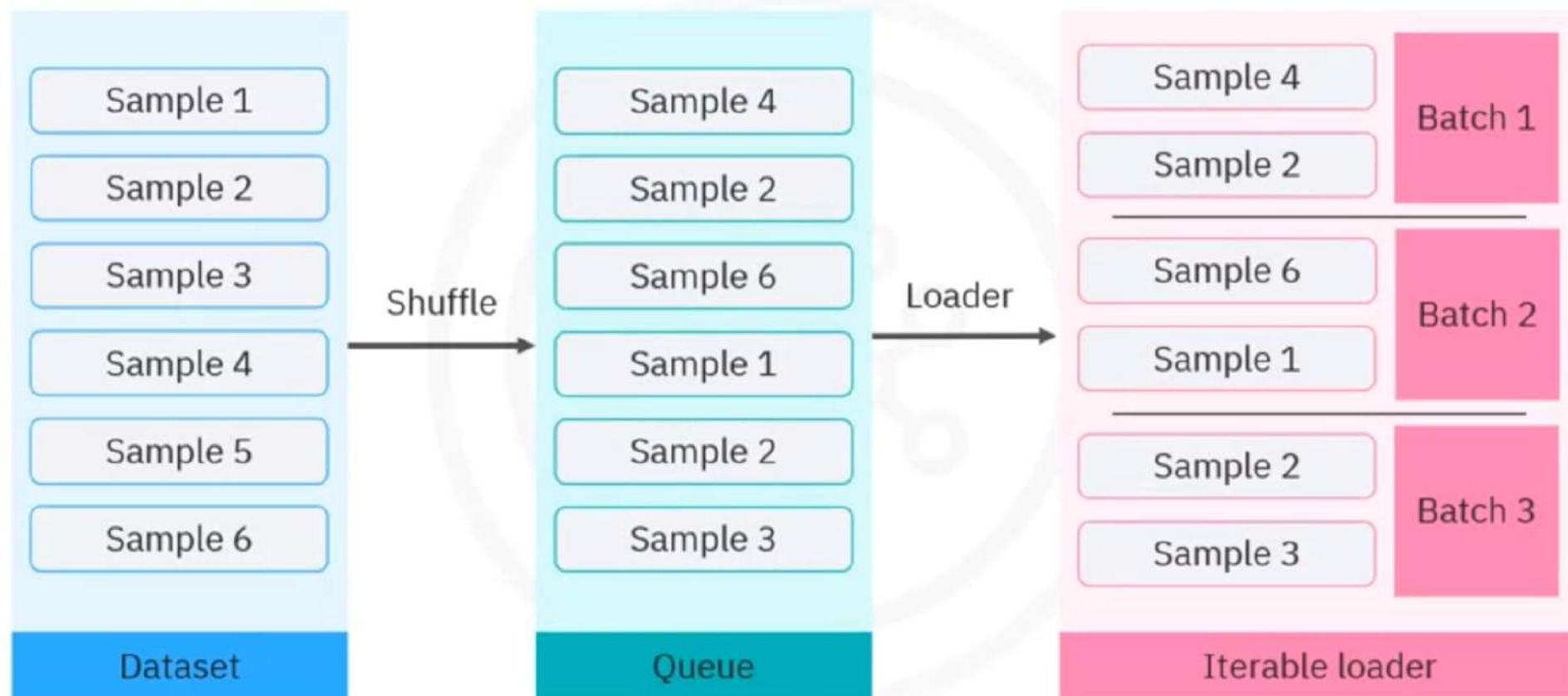
Sample 1

```
dataset[1]:
```

```
"It is our choices, Harry, that show what we truly are, far more than our  
abilities."
```

Sample 2

DataLoader: batch_size=2, shuffle=True



Iterator

- Iterator: Object that can be looped over
- Two methods: **iter()** and **next()**

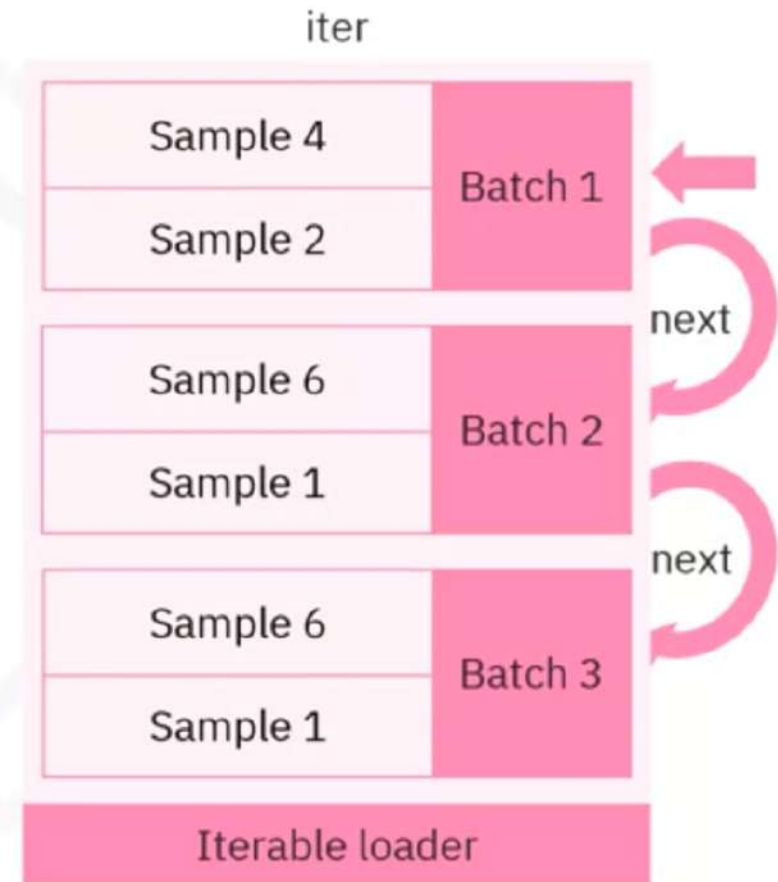
```
data_iter= iter(data_loader)
```

```
next(data_iter)
```

```
next(data_iter)
```

```
next(data_iter)
```

Sample 2	Batch 3
Sample 3	



Using iterator

```
from torch.utils.data import DataLoader

custom_dataset = CustomDataset(sentences)

batch_size = 2
dataloader = DataLoader(custom_dataset, batch_size=batch_size, shuffle=True)

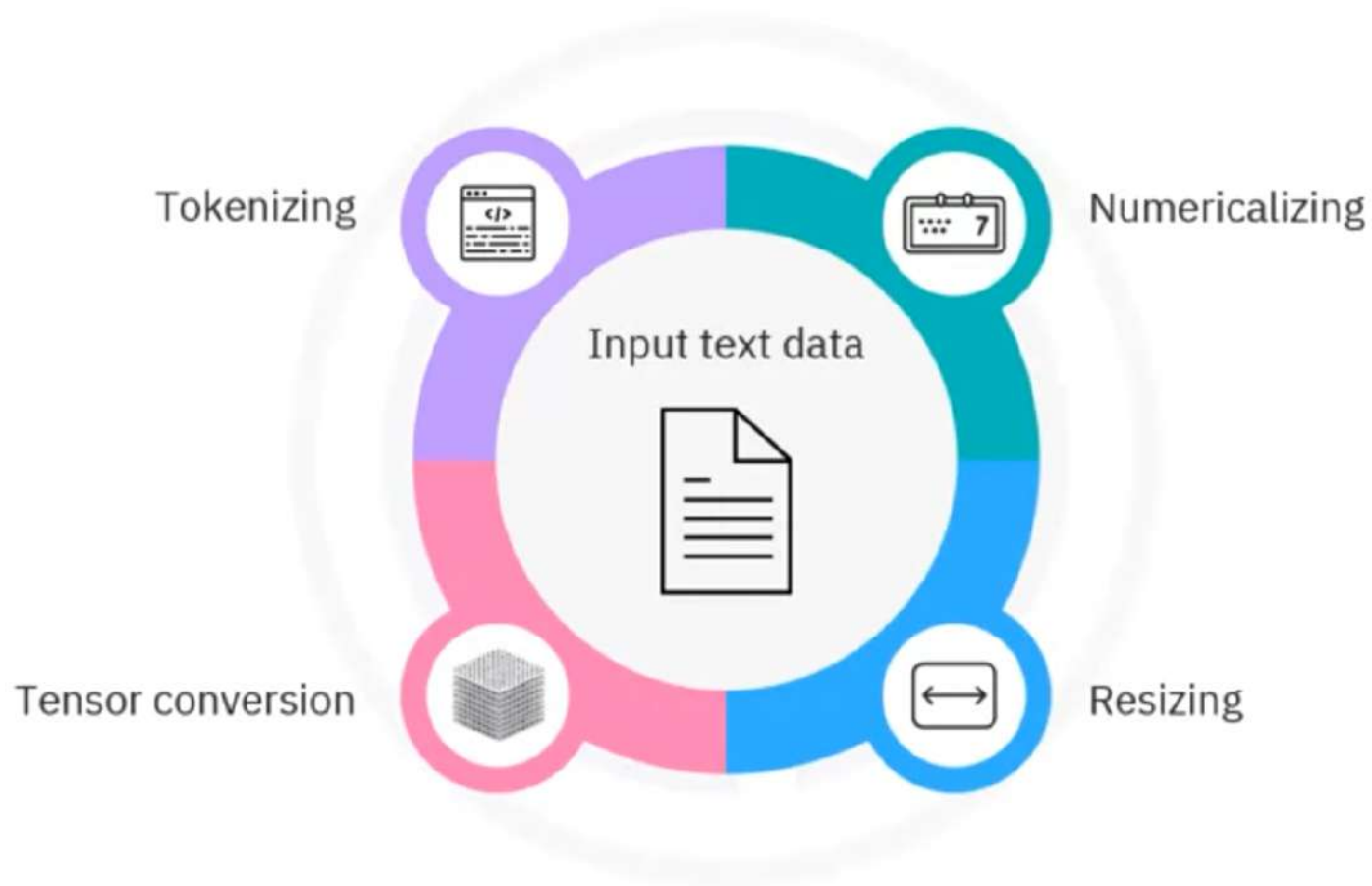
for batch in dataloader:
    print(batch)
```

OUTPUT:

```
['Soon we must all face the choice between what is right and what is easy.',
 'You are awesome!']
["Fae's a fickle friend, Harry.", 'It is our choices, Harry, that show what we
truly are, far more than our abilities.']
["If you want to know what a man's like, take a good look at how he treats his
inferiors, not his equals.", 'Youth cannot know how age thinks and feels. But
old men are guilty if they forget what it was to be young.']
```



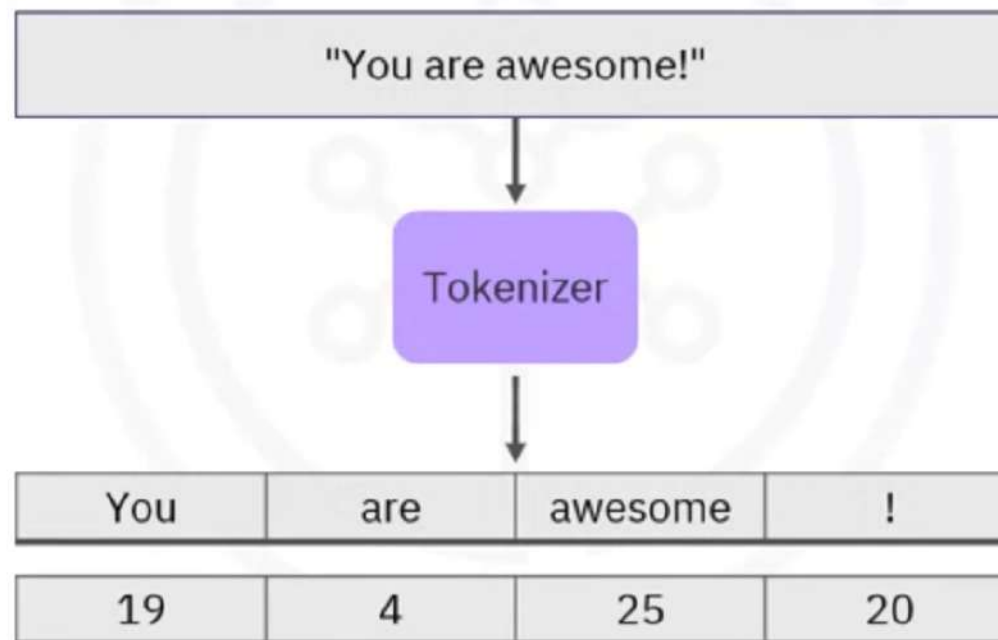
Transformation on input text data



Tokenization and vocabulary building

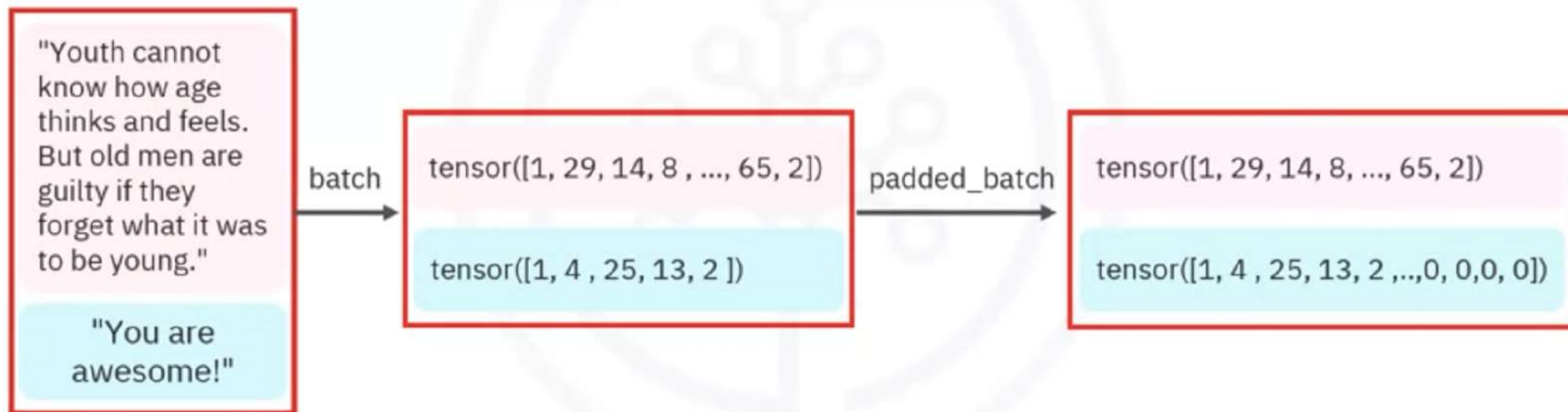
```
tokenizer = get_tokenizer("basic_english")
```

```
vocab = build_vocab_from_iterator(map(tokenizer, sentences))
```



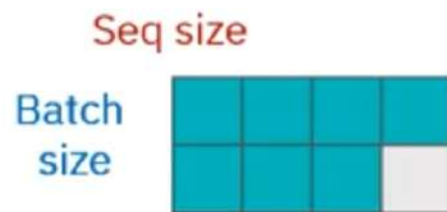
Handling variable-length data

```
from torch.nn.utils.rnn import pad_sequence  
for batch in dataloader:  
    padded_batch = pad_sequence(batch, batch_first=True, padding_value=0)
```

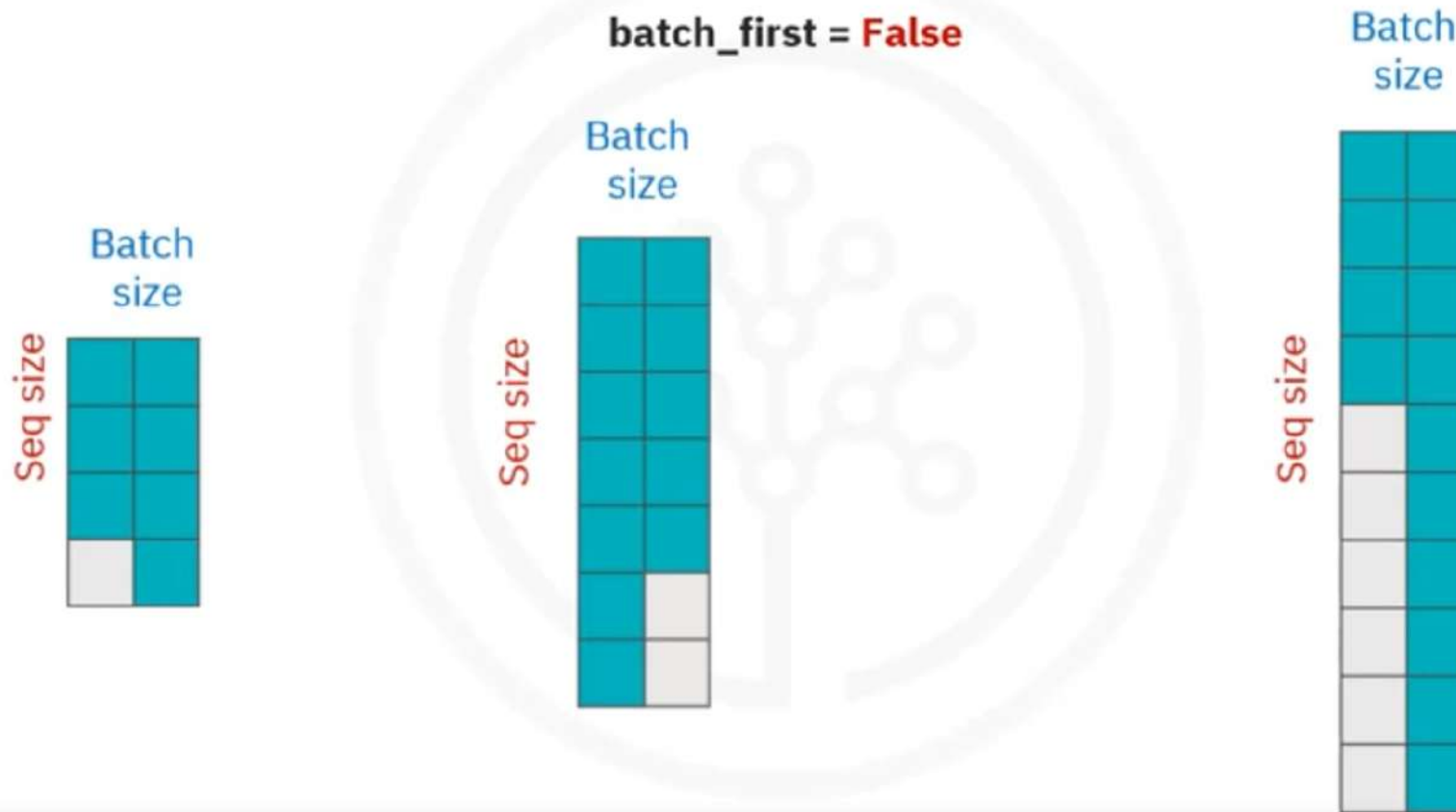


The batch_first argument

batch_first = True



The batch_first argument



batch_first = True

Fae's a fickle friend, Harry.

You are awesome!

<BOS> Fae 's a fickle
friend , Harry . <EOS>

<BOS> you are awesome ! <EOS> <PAD> <PAD>
<PAD> <PAD>

2-d tensor[B*S]

batch size(2)

<BOS> Fae 's a fickle friend , Harry . <EOS>

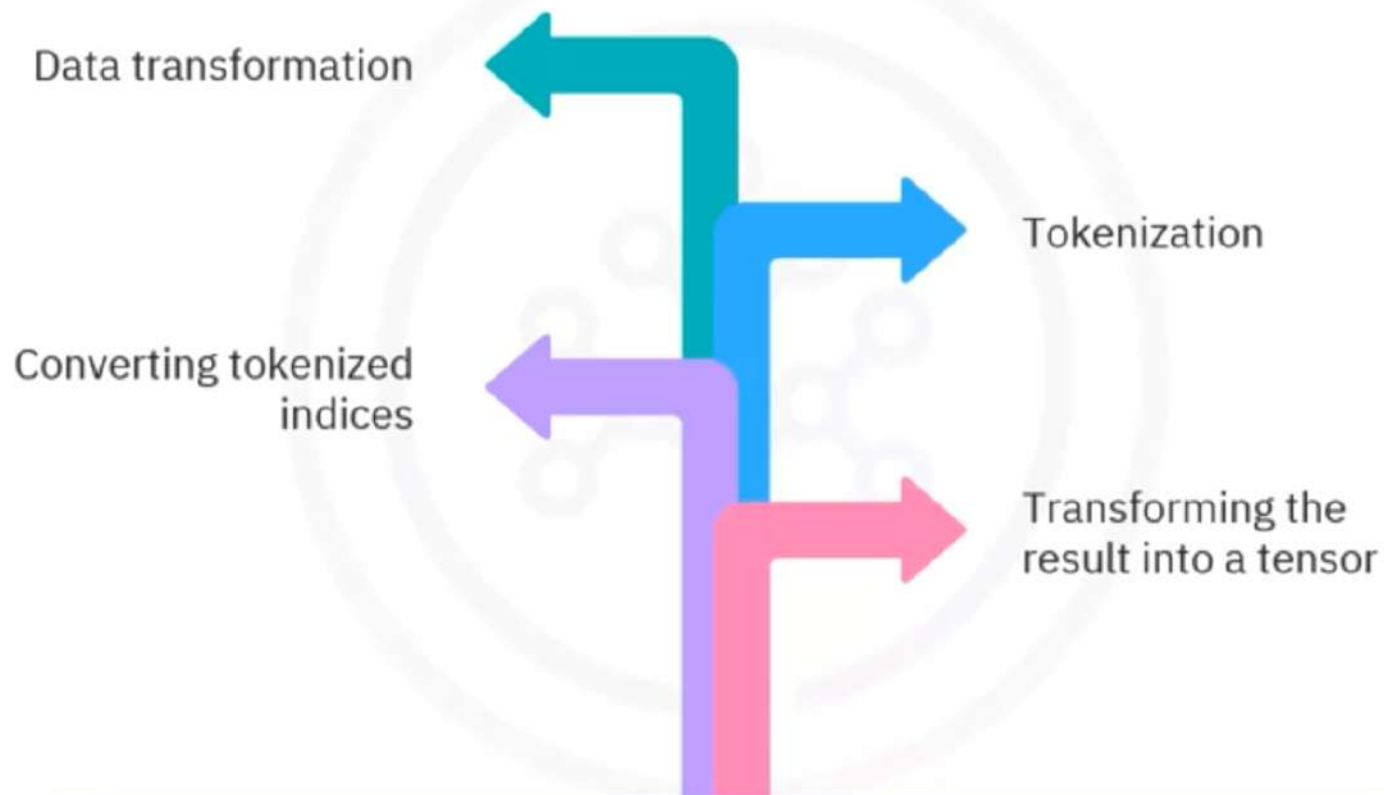
<BOS> you are awesome ! <EOS> <PAD> <PAD> <PAD> <PAD>

sequence length(10)

batch_first = False



Collate function



Collate function

```
def collate_fn(batch):  
    tensor_batch = []  
    for sample in batch:  
        tokens = tokenizer(sample)  
        tensor_batch.append(torch.tensor([vocab[token] for token in tokens]))  
    padded_batch = pad_sequence(tensor_batch, batch_first=True)  
    return padded_batch
```

```
dataloader = DataLoader(custom_dataset, batch_size=batch_size, shuffle=True,  
                        collate_fn=collate_fn)
```



Recap

In this video, you learned that:

- A data loader helps you prepare and load data to train generative AI models.
- PyTorch and TensorFlow have a dedicated `DataLoader` class.
- Data loaders enable efficient batching and shuffling of data and allow for on-the-fly processing.
- Data loaders seamlessly integrate with the PyTorch training pipeline and simplify data augmentation and preprocessing.
- Using data loaders, you can output data in batches instead of one sample at a time.

← Back Practice Quiz: Preparing Data
Practice Assignment • 10 min

🌐 English ▼

1. Which statement is true about the Unigram algorithm for tokenization?

1 point

- ☐ It evaluates the benefits and drawbacks of splitting and merging two symbols to ensure its decisions are valuable.
- ☐ It segments text into manageable parts and assigns unique IDs.
- ☒ It begins with a large list of possibilities and gradually narrows down based on how frequently they appear in the text.
- ☐ It involves splitting text into individual characters.

2. Identify the advantages of using data loaders in natural language processing (NLP). Select all that apply.

1 point

- ☒ Seamlessly integrates with the PyTorch training pipeline
- ☒ Enables batching of data
- ☒ Enables shuffling of data
- ☐ Splits text into characters to ensure vocabulary is small

3. Fill in the blank.

1 point

You can use the _____ to ensure that all sentences have the same length after tokenization, matching the length of the longest sentence among the input sentences.

- ☐ Underscore symbol
- ☐ <eos> special token
- ☐ ## symbol
- ☒ <pad> token

Activities Google Chrome 2 13:46

course.org/learn/generative-ai-llm-architecture-data-preparation/assignment-submission/NF5tU/graded-quiz-data-preparation-for-llms/attempt

Back Graded Quiz: Data Preparation for LLMs Graded Assignment • 15 min English Due Nov 6, 11:59 PM PST Finish update

1. Which tokenization method generates a smaller vocabulary but increases input dimensionality and computational needs? 1 point

- ☐ WordPiece tokenization
- ☐ SentencePiece tokenization
- ☐ Word-based tokenization
- ☒ Character-based tokenization

2. Imagine you are training a sentiment analysis model where the input consists of user reviews. After tokenization, you find that the sequences have varying lengths. Which concept will you employ to address the issue of varied lengths while using data loaders? 1 point

- ☐ Batching
- ☒ Padding
- ☐ Iteration
- ☐ Shuffling

3. Fill in the blank. 1 point

In subword-based tokenization, the _____ indicates that the word should be attached to the previous word without a space.

- ☐ <eos> special token
- ☐ Underscore symbol
- ☐ <pad> token
- ☒ ## symbol

Activities

Google Chrome

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Finish update

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Back

Graded Quiz: Data Preparation for LLMs

Graded Assignment • 15 min

English

Due Nov 6, 11:59 PM PST

☒ ## symbol

4. Identify an advantage of word-based tokenization. 1 point

☐ It creates smaller vocabulary

☒ It preserves the semantic meaning

☐ It evaluates the benefits and drawbacks of splitting and merging two symbols

☐ It breaks down infrequent words to meaningful subwords

5. Which input provided during data loader creation helps prevent the model from learning patterns based on the order of the data? 1 point

☐ The padding value

☐ The batch size

☐ The data set

☒ The shuffle argument

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You must select the checkbox in order to submit the assignment.

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