

1.3 PyTorch CORE MODULES

These comes along with pytorch as inbuilt

Core PyTorch Modules

Module	Description
<i>torch</i>	The core module providing multidimensional arrays (tensors) and mathematical operations on them.
<i>torch.autograd</i>	Automatic differentiation engine for recording operations on tensors to compute gradients for optimization.
<i>torch.nn</i>	Neural networks library for layers, activations, loss functions, and utilities for deep learning models.
<i>torch.optim</i>	Optimizers like SGD, Adam, RMSprop for training neural networks.
<i>torch.utils.data</i>	Data handling utilities including Dataset and DataLoader for efficient data management and loading.
<i>torch.jit</i>	Supports JIT compilation and TorchScript to optimize models and enable deployment without Python dependencies.
<i>torch.distributed</i>	Tools for distributed training across multiple GPUs/machines (parallel computation).
<i>torch.cuda</i>	Interfaces with NVIDIA CUDA for GPU acceleration in tensor computations and model training.
<i>torch.backends</i>	Settings/control over backend libraries (cuDNN, MKL, etc.) for performance tuning.
<i>torch.multiprocessing</i>	Utilities for parallelism with multiprocessing, supporting CUDA tensors.
<i>torch.quantization</i>	Tools for model quantization to reduce size and improve inference, especially on edge devices.
<i>torch.onnx</i>	Exporting PyTorch models to ONNX format for interoperability with other frameworks/deployment.

PyTorch Domain Libraries

Library	Description
<i>torchvision</i>	Datasets, model architectures, image transformations for computer vision tasks.
<i>torchtext</i>	NLP tools, datasets, data preprocessing, vocabulary management.
<i>torchaudio</i>	Audio processing utilities like I/O, transforms, pre-trained models for speech recognition.
<i>torcharrow</i>	Accelerated data loading/preprocessing for tabular/time series data (experimental).
<i>torchserve</i>	Model serving library for deploying trained models at scale in production environments.
<i>pytorch_lightning</i>	Lightweight wrapper for PyTorch, simplifies training loop and reduces boilerplate for scalable, reproducible models.

Popular PyTorch Ecosystem Libraries

Library	Description
<i>Hugging Face Transformers</i>	State-of-the-art pre-trained models for NLP tasks, translation, QA, built on PyTorch.
<i>Fastai</i>	High-level library for fast and accurate neural net training with modern best practices, built on PyTorch.
<i>PyTorch Geometric</i>	Extension for geometric deep learning (graph neural networks, 3D data processing).
<i>TorchMetrics</i>	Modular metrics API for standardized implementations, compatible with PyTorch Lightning.
<i>TorchElastic</i>	Dynamic scaling of PyTorch distributed training jobs; enables elasticity in resource management.
<i>Optuna</i>	Hyperparameter optimization framework, integrates well with PyTorch for tuning models.
<i>Catalyst</i>	High-level features for training neural networks, focusing on reproducibility and fast experimentation.
<i>Ignite</i>	Lightweight library for training/evaluating neural networks in PyTorch; helps with training workflows.
<i>AllenNLP</i>	Research library supporting deep learning for NLP, built on PyTorch.
<i>Skorch</i>	scikit-learn compatible wrapper, enables using PyTorch models with scikit-learn utilities/APIs.
<i>PyTorch Forecasting</i>	High-level library for time series forecasting; easy model building, training, evaluation.
<i>TensorBoard for PyTorch</i>	Visualization of training metrics, model graphs, and data within TensorBoard for PyTorch.