### Intensive 16-week program curriculum

#### Week 01 - Introduction

- → Overview of the curriculum
- → Important roles to pursue in the future
  - ◆ Conventional Routes
    - Data Engineer
    - Machine Learning (ML) Engineer
    - Data Analyst
    - Data Scientist
    - Research Engineer
    - Research Scientist
    - Academic Researcher
  - ◆ Unconventional Routes
    - Freelance Data Science Engineer
    - Data Science Trainer or Mentor
    - Entrepreneur
- → A day in the life (From Experience)
  - ◆ ML Engineer
  - ◆ Data Scientist
- → Guide to Discipline (From Experience)
  - ◆ Limiting beliefs and how to fix them
  - ◆ Common reasons for failure
  - ♦ From Motivation to Discipline

#### Week 02 - Python Specialization

- → Reviewing Basic Programming Concepts
  - **♦** Time Complexity
  - **♦** Sorting
  - ◆ Greedy Algorithms
  - ◆ Dynamic Programming
- → Set Up Tools and Environments for Data Science
  - ◆ Virtual environments and why we need them
  - ◆ VS Code
  - ♦ Jupyter Notebook
  - ◆ Google Colaboratory
  - ◆ Kaggle
- → How to Python
  - **♦** Script Structures
  - ◆ Conditionals and Loopings
  - **♦** Functions
  - ◆ Objects and Classes
  - ◆ Data Structures
  - **♦** Comprehensions
  - ◆ Debugging and Error Handling
- → How to NumPy
  - ♦ Why use NumPy?
  - ♦ NumPy Basics
  - ♦ Array Manipulation

### Week 03 - Data ETL (Extract, Transform, Load)

- → Understanding End-to-end Data Science Pipeline
  - ◆ Data ETL
  - ◆ Data Analytics and Visualization
  - ◆ ML Modeling
  - ◆ Deployment and Performance Tracking
- → Crawl a website
  - ♦ Why?
  - ♦ How? Full Demonstration
- → Using Pandas for ETL
  - ♦ Pandas Demo
    - Basics
    - Advanced Pandas Tricks
  - ◆ Data Manipulation
  - ♦ Quantitative Data Analysis

### Week 04 - Data Visualization

- → Visualization with Python
  - ◆ Matplotlib
  - ◆ Seaborn
  - ◆ Plotly
- → Intro to Emerging Tools
  - ◆ Tableau
  - ◆ StreamLit

# Week 05 - Data Analytics with Tableau

- → Tableau Competitors
  - ◆ Microsoft Excel
  - **♦** KNIME
  - ◆ Power BI
- → Why Tableau?
- → Tableau Setup and ETL
- → Filters
- → Charts and Graphs
- → Dashboards
- → Parameters, Functions
- → Blending, Joining

# Week 06 - Data Analytics Capstone Project

A project with a particular problem statement will be assigned to judge the following skills

- → Basic Web Scraping
- → Data Manipulation
- → Data Analysis
- → Data Visualization
- → Ability to use Tableau and StreamLit

### Week 07 - Introduction to Deep Learning

- → Origins to SOTA (State of the Art)
- → Introducing necessary software
  - ◆ PyTorch
  - ◆ fastai
  - ◆ HuggingFace
  - **♦** Gradio
  - ♦ Weights and Biases
- → Transfer Learning and Finetuning
- → Train our first Computer Vision (CV) model
- → Machine Learning (ML) and its types
- → When to use ML
- → How our CV model works
- → What our CV model learns
- → Train our first Segmentation, Tabular, and Recsys model
- → Importance of Validation and Test Sets

# Week 08 - End-to-end Data Science Pipeline in Practice

- → Data Collection
- → Data to DataLoaders
- → Data Augmentation
- → Training a model and Cleaning Data
- → Deploy our first model
- → Performance Monitoring
- → Multi-label classification
- → Regression

### Week 09 - Image Recognizer Project and Data Ethics

- → Individual Project: Build your own image recognizer with fastai from data gathering to model deployment
- → Convolutional Neural Networks
- → Residual Neural Networks
- → Data Ethics
  - ◆ Examples of ethical disasters
  - ♦ The necessity of Data Ethics
  - ◆ Recourse and Accountability
  - ◆ Feedback Loops
  - ◆ Bias
  - **♦** Disinformation

### Week 10 - Deep Dive into Deep Learning Foundations

- → Building a baseline
- → Gradient Descent
- → Loss Functions
- → Optimizer, Momentum, RMSProp, Adam
- → Nonlinearities
- → Building a PyTorch model from scratch
- → Learning Rate Finder
- → Discriminative Learning Rates
- → Overfitting, extrapolation problems
- → Exploding and vanishing gradient problems
- → Propose team and project ideas for the final capstone project

# Week 11 - Advanced Deep Learning Tricks

- → Normalization
  - ◆ Batch Normalization
  - ◆ Layer Normalization
- → Progressive Resizing
- → Test Time Augmentation
- → Mixup
- → Label Smoothing
- → Weight Decay
- → Final capstone project discussion

### Week 12 - Collaborative Filtering and Tabular Modeling

- → Collaborative Filtering (Collab)
  - ◆ Where we need this
  - ◆ Processing the data
  - ◆ Collab model with PyTorch
  - ◆ Collab model with fastai
  - ◆ Interpreting the results
  - ◆ Custom Collab DL model
- → Tabular Modeling
  - ♦ Processing the data
  - ◆ Decision Trees
  - **♦** Random Forests
  - ◆ Feature Importance
  - ◆ Data Leakage
- → Final capstone project discussion

### Week 13 - Natural Language Processing (NLP)

- → Preprocessing Text
- → Self-supervised learning
- → RNN, LSTM
- → Language Models with PyTorch and fastai
- → Text Classifier with PyTorch and fastai
- → Text Generation
- → Risks of Language Models
- → Transformers
- → HuggingFace
- → Final capstone project discussion

#### Week 14 - Experiment Management, Model Deployment, and Monitoring

- → Experiment Management
  - ◆ Why do we need it?
  - ◆ EDA with Weights and Biases (W&B)
  - ◆ Project Management, Artifact Versioning
  - ◆ Collaboration tools
  - ◆ Hyperparameter sweeps
- → Model Deployment
  - ◆ Batch Prediction
  - ◆ Rest APIs
  - ◆ Performance Optimization
  - ◆ Horizontal Scaling
  - ◆ Edge Prediction
- → Model Monitoring
  - ♦ Why do we need it?
  - ♦ Domain Shift and Data Drift
  - ♦ What and How to Monitor
  - ◆ Tools for monitoring
- → Final capstone project discussion

### Week 15 and 16 - Deep Learning Capstone Project

A project of medium difficulty upon discussion with the course instructor. The project should include the following:

- → Data Gathering, Processing, Cleaning
- → Maintain the Data Ethics
- → Building Models with PyTorch and/or fastai
- → Utilizing some advanced deep learning tricks
- → Experiment management with Weights and Biases
- → Model deployment and monitoring