

# **ML Engineer Program Syllabus**

#### **About FourthBrain**

FourthBrain educates aspiring Machine Learning engineers on the technical and practical skills required to provide immediate value to an AI product development or R&D team. Our programs emphasize open collaboration and communication, and our unique approach to cohort-based learning is aimed at helping you achieve your personal career goals while giving you the best chance to develop strong, lasting relationships with other ML practitioners on similar journeys.

### **Your Learning Transformation**

By the end of this course, you will be able to contribute to high-performing Al product teams by leveraging real-world data to **build**, **package**, and **deploy** state-of-the-art ML models as containerized web applications in cloud-based production environments.

#### **Capstone Project**

Capstone projects are designed to demonstrate your understanding of MLE software development and its implications. This includes understanding the potential business-value of your application and its extensibility, as well as the tools required to build, optimize, package, and deploy your ML models in a production software development environment. Projects are typically developed in groups of two or three; solo projects are possible, especially in situations where external support (i.e., your colleagues or other collaborators are available). Final deliverables will include a final presentation and GitHub repo, which together will allow you to share the details of what you've accomplished with your potential users, collaborators, employers, or the wider open-source ML community. You can see past capstone projects <a href="https://example.com/here-collaborators">here</a>.

#### **Career Growth**

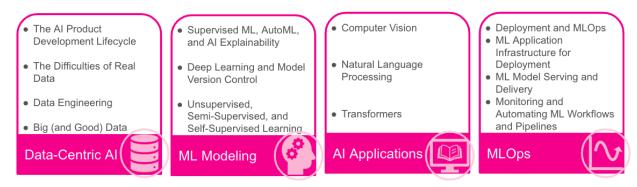
The Machine Learning Engineer program is designed for you to acquire the skills and knowledge required to work on a Machine Learning Engineering team. Whatever your reason for taking the program - to get a new role at a new company, to gain skills for your current job, or just for fun - we will support your career growth by helping you connect to professionals and employers, via guest speaking events and inviting employers to the final project presentation day. Career services assistance is available during the program and after graduation to help ensure that all candidates achieve their career goals.



## **Program Format**

Each week, we will curate content related to the weekly topic for you to review asynchronously. You will come together with your instructors and peers to discuss, contextualize, and practice the concepts for the week.

The Machine Learning Engineer program is divided into four pillars: Data Centric Al, Machine Learning Modeling, Al Applications, and MLOps.



During our first week we kick off the course with a review of important ML software engineering tools. We get our development environment set up and test out that we can build basic ML models.

Our last week is completely focused on Demo Day for your projects!



# Weekly Schedule

| Week   | Live Session<br>Concepts  | Coding<br>Exercise(s)  | Capstone   |
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| Week 1 Saturday Dec. 3 Getting Started with ML Engineering | <ul> <li>Program Overview</li> <li>Software         Development         Environment setup     </li> </ul>   | <ul> <li>MLE Basic <u>Tools</u>:</li> <li>Unix CLI, Git,<br/>Conda, Pip,<br/>Jupyter</li> <li>Pandas,<br/>Matplotlib,<br/>Seaborn, Sci-Kit<br/>Learn</li> </ul>  | Networking   |
| Week 2 Saturday Dec. 10 Al Product Development             | <ul> <li>Al Product         Development         Lifecycle &amp; ML         Project Scoping         Data Centric Al         Responsible ML         Principles     </li> </ul>  | <ul> <li>Creating a sentiment analyzer</li> <li><u>Tools</u>: Hugging Face Transformers Library, Twitter API</li> </ul>  | <ul> <li>Ideation         Workshop</li> <li>Al Product/ML         Project Scoping</li> </ul> |
| Week 3 Saturday Dec.17 Building ML App Prototypes          | <ul> <li>Ensuring High         Quality Data</li> <li>Fine-Tuning         Pre-Trained         Models</li> <li>Building ML         Proofs-of-Conce         pt (POCs) and         Minimum Viable         Products (MVPs)</li> <li>Application         Programming         Interfaces (APIs)</li> </ul> | <ul> <li>Collect real data from Twitter and Reddit APIs</li> <li>Fine-Tune a pre-trained transformer for sentiment analysis</li> <li>Develop a data-centric proof of concept</li> <li>Tools: Hugging Face Trainer API, Reddit API</li> </ul> | Peer Review and Teaming  |

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| Break   | Winter Holidays   |   |  |
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| Week 4, Saturday Jan. 7  Data Engineering                         | <ul> <li>Data Engineering         Workflows</li> <li>Data Wrangling &amp;         Exploratory Data         Analysis</li> <li>Feature Selection &amp;         Engineering</li> <li>Data Leakage</li> <li>Building ETL         Pipelines</li> </ul> | <ul> <li>Exploring and wrangling structured data to predict sales with simple ML pipeline</li> <li>Track and manage datasets</li> <li>Build an ETL workflow</li> <li>Tools: Airflow, DVC</li> </ul>                   | <ul> <li>Project         Proposals     </li> <li>Project Pitch Day         I     </li> </ul> |
| Week 5 Saturday Jan. 14 Big (and Good) Data                       | <ul> <li>Types of Distributed<br/>Computing for ML</li> <li>Data Preparation</li> <li>Big Data Tool<br/>Landscape</li> <li>How Good Data<br/>Becomes Big Data</li> </ul>  | <ul> <li>Build an ML pipeline and perform distributed hyperparameter tuning to predict subscriptions &amp; promotions</li> <li>Tools: Spark, Delta Lake, MLlib</li> </ul>   | Exploratory Data<br>Analysis and<br>Data Lineage<br>Documentation                            |
| Week 6 Saturday Jan. 21 Supervised ML, AutoML, and Explainability | <ul> <li>Essential         Regression &amp;         Classification         Algorithms</li> <li>Data Imbalance</li> <li>Accuracy Metrics</li> <li>AutoML Libraries</li> <li>Al Explainability</li> </ul>   | <ul> <li>Detect data imbalances and predict electronics purchases with explainable ML pipelines</li> <li>Perform AutoML to search for an optimal tree-based pipeline</li> <li>Tools: TPOT, SHAP, Streamlit</li> </ul> | Establish ML     Modeling     Baseline   |



| Week 7 Saturday Jan. 28  Deep Learning and Model Version Control                    | <ul> <li>Neural Network         Basics</li> <li>Recurrent Neural         Networks (RNNs)</li> <li>Long Short-Term         Memory Networks         (LSTMs)</li> <li>Generative         Adversarial         Networks (GANs)</li> <li>ML Model Version         Control</li> </ul> | <ul> <li>Build, train, and evaluate a neural network for fuel efficiency prediction from scratch</li> <li>Compare neural network performance based on type</li> <li>Tools: MLflow Tracking</li> </ul>   | Initial GitHub     Documentation |
|---|--|---|----------------------------------|
| Week 8 Saturday Feb. 4  Unsupervised, Semi-Supervised, and Self-Supervised Learning | <ul> <li>Dealing with         Unstructured Data</li> <li>Clustering</li> <li>Dimensionality         Reduction</li> <li>Label         propagation/label         spreading</li> <li>Co-training         algorithms</li> <li>Zero-shot learning</li> </ul>                        | <ul> <li>Performing         Customer         Segmentation</li> <li>Predicting         product sales         using         semi-supervised         learning</li> <li>Tools: Sk-learn         (pca, Kmeans,         Silhouette         Analysis)</li> </ul> | Data & Model<br>Iterations       |
| Week 9 Saturday Feb. 11 Computer Vision   | <ul> <li>Convolutional<br/>Neural Networks<br/>(CNNs)</li> <li>Computer Vision<br/>Benchmarks</li> <li>Dealing with Images</li> <li>Object Detectors</li> <li>Semantic<br/>Segmentation</li> <li>Explainability &amp;<br/>Saliency</li> </ul>                                  | <ul> <li>Few-shot dog detection dogs through fine-tuning of a pre-trained single shot multibox detector</li> <li>Tools:         <ul> <li>TensorFlow</li> <li>Model Garden</li> </ul> </li> </ul>  |                                  |



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| Week 10<br>Saturday Feb. 18<br>Natural Language<br>Processing                                | <ul> <li>Recurrent Neural         Networks and Long         Short-Term Memory         Networks (RNNs,         LSTMs))</li> <li>Natural Language         Benchmarks</li> <li>Dealing with Text</li> <li>Tokenization &amp;         Word Embeddings</li> <li>Bag of Words</li> <li>Large Language         Models (LLMs)</li> </ul> | <ul> <li>Analyze tweets using basic NLP tasks</li> <li>Build hate speech detectors using a naive Bayes classifier and a bidirectional LSTM</li> <li>Tools:         <ul> <li>Tensorboard Embedding Projector, LIME Text Explainer</li> </ul> </li> </ul> | Project Pitches II  |
| Week 11<br>Saturday Feb. 25<br>Foundation<br>Models:<br>Transformers and<br>Diffusion Models | <ul> <li>Encoder and Decoder Networks</li> <li>General Pre-Trained Transformers (e.g., GPT-3)</li> <li>Fine-Tuning Foundation Models</li> <li>Diffusion Models (e.g., DALL·E 2)</li> </ul>   | <ul> <li>Build a transformer model for news article text classification from scratch</li> <li>Fine-tune a pre trained transformer</li> <li>Tools: Hugging Face Models</li> </ul>  | GitHub     Documentation     Revisions  |
| Week 12<br>Saturday Mar. 4<br>Manual MLOps<br>Deployments                                    | <ul> <li>MLOps Level 0:         Manual</li> <li>Intro to VS Code</li> <li>Branch         Development</li> <li>AWS (Browser)</li> <li>AWS via VS Code</li> <li>FastAPI &amp; Web App         Frameworks</li> <li>File Structure for         Web Applications</li> </ul>   | <ul> <li>Deploy a public<br/>API stock<br/>prediction<br/>service</li> <li><u>Tools</u>: VS Code,<br/>Git, AWS,<br/>FastAPI</li> </ul>  | <ul> <li>Deploy your ML<br/>Application<br/>Locally</li> <li>Deploy your ML<br/>App to AWS</li> </ul> |



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| Week 13 Saturday Mar. 11 ML Application Infrastructure for Deployment | <ul> <li>Fast API         Architecture         (Swagger UI,         Uvicorn)</li> <li>Def: Deploy,         Endpoint, Ports</li> <li>Review API/HTTP</li> </ul>  | <ul> <li>Build and deploy a containerized stock prediction service</li> <li>Manage model versions with a model registry</li> <li>Tools: Docker, MLflow Model Registry</li> </ul> | <ul> <li>Containerize and deploy your ML web application and deploy locally and to AWS</li> <li>Infrastructure Diagramming</li> </ul>                                       |
| Week 14 Saturday Mar. 18 Containers                                   | <ul> <li>OS and VM Basics</li> <li>Containers and<br/>Container<br/>Orchestration</li> <li>Model Management<br/>and Model<br/>Registries</li> </ul>   | <ul> <li>Implement stock prediction app within a serverless computing service</li> <li>Deploy stock prediction service</li> <li>Tools: AWS Lambda, API Gateway</li> </ul>        | <ul> <li>1-Page Narrative</li> <li>Final GitHub         Documentation             Revisions     </li> <li>Last-Minute         Feature             Additions     </li> </ul> |
| Week 15<br>Saturday Mar. 25<br>MLOps Pipeline<br>Monitoring           | <ul> <li>ML Metric         Monitoring</li> <li>Ops Monitoring</li> <li>Logging and the         ELK Stack</li> <li>Observability</li> <li>MLOps L1: Pipeline         Automation &amp;         Retraining Triggers</li> </ul> | <ul> <li>Tools for detecting monitoring drift in Diabetes prediction</li> <li>Tools: Prometheus, Grafana</li> </ul>  | <ul> <li>Elevator Pitch</li> <li>Final Capstone         Presentation and             Live Demo             Practice and             Submission     </li> </ul>              |
| Week 16<br>Saturday Apr. 1<br>Demo Week!                              | <ul><li>Storytelling<br/>Workshop</li><li>Capstone Practice</li></ul>   | <ul> <li>Final Capstone<br/>GitHub<br/>submission!</li> </ul>  | Demo Day!   |