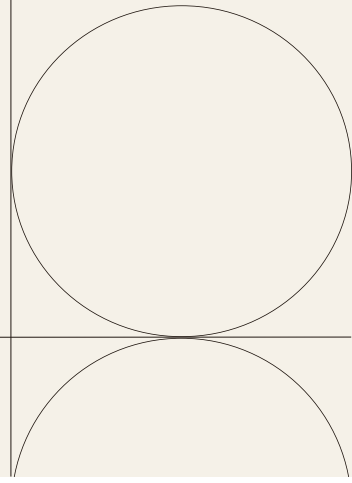


AYOMI AI - Classes



DATA Science (Syllabus)

1. Introduction to Data Science	<ul style="list-style-type: none"> • What is Data Science? • Data Science vs. Machine Learning vs. Artificial Intelligence • Applications of Data Science • Roles and Responsibilities of a Data Scientist
2. Mathematics and Statistics for Data Science	<ul style="list-style-type: none"> • Linear Algebra (vectors, matrices, eigenvalues/eigenvectors) • Probability Theory (distributions, Bayes' theorem) • Descriptive Statistics (mean, median, variance, etc.) • Inferential Statistics (hypothesis testing, p-values, confidence intervals) • Sampling Methods and Estimations • Regression Analysis (simple and multiple)
3. Programming for Data Science	<ul style="list-style-type: none"> • Introduction to Python or R <ul style="list-style-type: none"> ◦ Python Libraries: NumPy, Pandas, Matplotlib, Seaborn ◦ R Libraries: ggplot2, dplyr, tidyr • Data Structures (lists, arrays, dataframes) • Functions, Loops, and Conditional Statements • File Handling (reading, writing data from CSV, JSON, SQL)
4. Data Preprocessing	<ul style="list-style-type: none"> • Data Cleaning (handling missing data, outliers) • Data Transformation (scaling, encoding, normalization) • Feature Engineering (creating new features, feature selection) • Text Data Processing (tokenization, stopwords, TF-IDF)
5. Exploratory Data Analysis (EDA)	<ul style="list-style-type: none"> • Data Visualization (histograms, boxplots, scatterplots, etc.) • Correlation and Covariance • Identifying trends and patterns • Summary Statistics • Using Matplotlib, Seaborn, and Plotly for visualizations
6. Machine Learning	<ul style="list-style-type: none"> • Supervised Learning: <ul style="list-style-type: none"> ◦ Linear Regression ◦ Logistic Regression ◦ Decision Trees and Random Forests ◦ Support Vector Machines (SVM) ◦ k-Nearest Neighbors (k-NN) ◦ Neural Networks

	<ul style="list-style-type: none"> ● Unsupervised Learning: <ul style="list-style-type: none"> ○ Clustering (K-means, Hierarchical) ○ Dimensionality Reduction (PCA, t-SNE) ● Model Evaluation (cross-validation, metrics like accuracy, precision, recall, F1 score)
7. Deep Learning	<ul style="list-style-type: none"> ● Introduction to Neural Networks ● Backpropagation and Gradient Descent ● Convolutional Neural Networks (CNNs) ● Recurrent Neural Networks (RNNs) and LSTMs ● TensorFlow and Keras
8. Natural Language Processing (NLP)	<ul style="list-style-type: none"> ● Text Preprocessing (tokenization, lemmatization, stemming) ● Bag-of-Words, TF-IDF, Word2Vec ● Sentiment Analysis ● Named Entity Recognition (NER) ● Text Classification and Clustering
9. Big Data and Data Engineering	<ul style="list-style-type: none"> ● Introduction to Big Data Tools (Hadoop, Spark) ● Distributed Data Processing ● Working with NoSQL Databases (MongoDB, Cassandra) ● Data Pipelines and ETL (Extract, Transform, Load) ● Cloud Platforms (AWS, Azure, Google Cloud)
10. Model Deployment and Production	<ul style="list-style-type: none"> ● Model Deployment using Flask/Django ● Introduction to REST APIs ● Cloud Deployment (AWS, GCP, Heroku) ● Continuous Integration and Continuous Deployment (CI/CD) ● Model Monitoring and Maintenance
11. Capstone Project	<ul style="list-style-type: none"> ● Working on a real-world dataset ● Applying Data Science techniques to solve a business problem ● Presenting results and insights using visualizations
Optional: Special Topics in Data Science	<ul style="list-style-type: none"> ● Reinforcement Learning ● Advanced Deep Learning (GANs, Transformers) ● Time Series Analysis ● Recommender Systems

