



DATA SCIENTIST

 **LIVE COURSE**

Join us for an insightful journey into Data Science!



March 1, 2025



Start at 5 PM IST

**Ready to level up your data skills?
Reserve your spot now!!**

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ABOUT US



Founder @ Euphoria State
AI/ML Innovator
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Co-Founder @ Euphoria State
AI Chip Designer @ Intel
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Co-Founder @ Euphoria State
Prev. Product Head @
CloudyML
[Linkedin](#)

Euphoria state is born with a simple vision, to make quality education accessible to everyone , no matter their background and provide right guidance and support to achieve greatness.



Curriculum

Learn step-by-step with weekly live classes, practical assignments, and hands-on projects.



Week 1: Python Basics and Foundational Concepts

- Introduction to Python
 - Syntax, indentation, comments, understanding Python's interpreter
- Variable Naming and Datatypes
 - Naming conventions, numeric types (int, float, complex), strings, type conversion, type hints
- Basic Input/Output
 - Using input() and print(), string formatting (f-strings, .format())

Week 2: Control Flow and Collections

- Control Flow
 - Conditional statements (if-else, nested conditions), logical operators (and, or, not)
 - Loops: for and while loops, break, continue, else with loops
- Data Structures: Lists and Tuples
 - Creating lists, list indexing, slicing, list methods (append, pop, extend, sort)
 - Tuples: Immutable properties, tuple unpacking, practical use cases



Week 3: Advanced Data Structures and Comprehensions

- Dictionaries
 - Creating dictionaries, accessing/updating/deleting keys, dictionary methods (keys(), values(), items())
 - Nested dictionaries, practical examples (JSON-like data)
- Sets
 - Creating sets, set operations (union, intersection, difference, symmetric_difference)
- List/Dictionary Comprehensions
 - Using comprehensions to create lists and dictionaries efficiently, conditional comprehensions

Week 4: Functions and Modules

- Functions
 - Defining functions, parameters and arguments, default arguments, return statements
 - Anonymous functions (lambda), function annotations, scope (local, global, nonlocal)
- Modules and Packages
 - Importing modules (import, from ... import), using standard libraries (math, random, os)
 - Working with custom modules and `__init__.py`



Week 5: File Handling and Error Management

- File Handling
 - Reading and writing text files (`open()`, `read()`, `write()`, file modes `r`, `w`, `a`)
 - Working with CSV and JSON files (`csv` module, `json` module)
- Error and Exception Handling
 - Understanding exceptions, using `try`, `except`, `finally`, raising custom exceptions (`raise`)

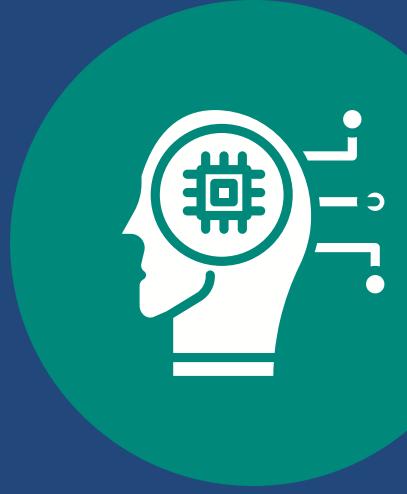
Week 6: Python for Data Analysis

- Introduction to NumPy
 - Arrays vs. lists, creating arrays, array operations, broadcasting, indexing/slicing, reshaping arrays
- Introduction to pandas
 - Series and DataFrames, creating DataFrames, basic operations (`head()`, `describe()`, `info()`), indexing and slicing and other methods
- Basic Data Visualization
 - Introduction to matplotlib and seaborn, creating line plots, bar charts, and scatter plots

Week 7-8: Advanced Python Concepts and Project

- Object-Oriented Programming (OOP)
 - Classes and objects, attributes and methods, inheritance, polymorphism
- Python Project
 - Exploratory Data Analysis (EDA) project using pandas, NumPy, matplotlib, and seaborn

MACHINE LEARNING

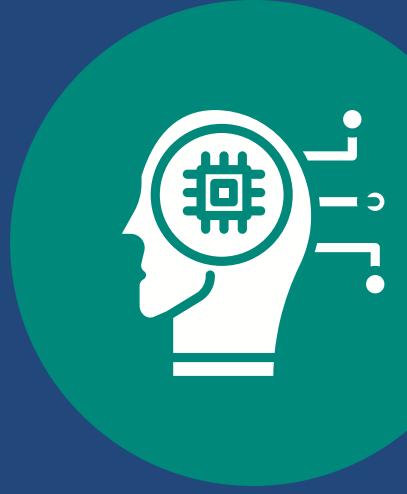


Week 1: Introduction to Machine Learning and Data Preprocessing

- Introduction to Machine Learning
 - Definition and importance, types of ML (supervised, unsupervised, reinforcement learning)
 - Real-world applications: recommendation systems, fraud detection, sentiment analysis
- ML Workflow
 - Problem definition, data collection, preprocessing, modeling, evaluation, and deployment
- Data Preprocessing
 - Handling missing data (mean/mode/median imputation, forward/backward fill)
 - Encoding categorical variables (one-hot encoding, label encoding)
 - Feature scaling (min-max scaling, standardization, normalization)
- Exploratory Data Analysis (EDA)
 - Identifying outliers, correlation analysis, visualizing distributions



MACHINE LEARNING

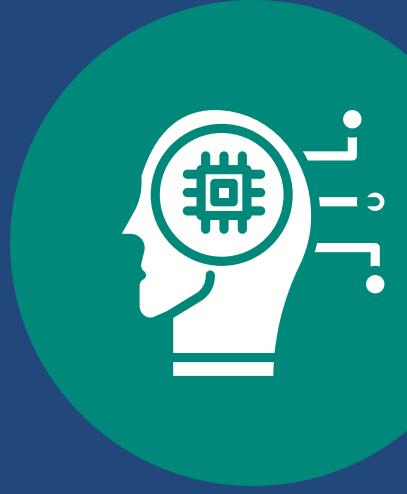


Week 2: Supervised Learning - Regression

- Introduction to Regression
 - Regression vs. classification, when to use regression models
- Linear Regression
 - Mathematical intuition, assumptions, cost function (MSE), gradient descent
 - Implementation with scikit-learn (LinearRegression), interpreting coefficients
- Polynomial Regression
 - When linearity doesn't fit, polynomial feature transformation, overfitting and underfitting
- Regularization Techniques
 - Ridge regression (L2 regularization), Lasso regression (L1 regularization)
- Model Evaluation Metrics
 - Mean Absolute Error (MAE), Mean Squared Error (MSE), R² score



MACHINE LEARNING

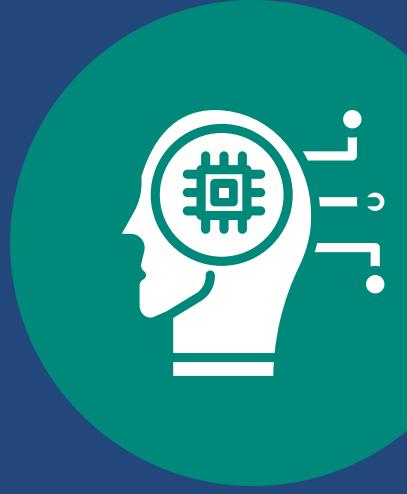


Week 3: Supervised Learning - Classification

- Introduction to Classification
 - Binary vs. multi-class classification, decision boundaries
- Logistic Regression
 - Mathematical intuition, sigmoid function, decision threshold tuning
 - Practical implementation using scikit-learn
- K-Nearest Neighbors (KNN)
 - Distance metrics (Euclidean, Manhattan), choosing k, bias-variance tradeoff
- Decision Trees
 - Splitting criteria (Gini impurity, entropy), visualizing decision trees, pruning
- Random Forests
 - Bagging, Boosting, feature importance, handling overfitting with ensemble methods
- Model Evaluation Metrics for Classification
 - Confusion matrix, precision, recall, F1-score, ROC-AUC



MACHINE LEARNING

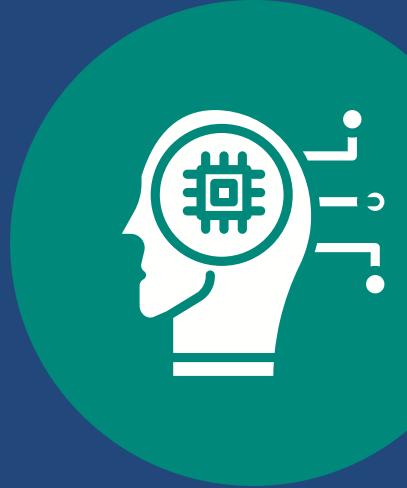


Week 4: Unsupervised Learning and Advanced Topics

- Clustering Techniques
 - K-Means: Centroid initialization, elbow method for determining k
 - Hierarchical Clustering: Agglomerative clustering, dendograms
 - DBSCAN: Density-based clustering, epsilon and min_samples
- Dimensionality Reduction
 - PCA (Principal Component Analysis): Explained variance, feature transformation
 - t-SNE (t-Distributed Stochastic Neighbor Embedding): Visualization of high-dimensional data
- Anomaly Detection
 - Techniques for outlier detection, applications in fraud detection and system monitoring



MACHINE LEARNING



Week 5: Model Optimization and Deployment

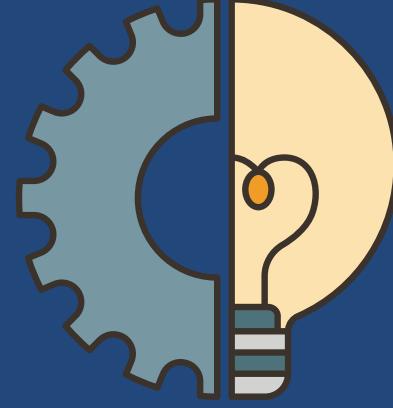
- Model Optimization
 - Cross-validation: K-fold, leave-one-out, stratified K-fold
 - Hyperparameter Tuning: Grid search, random search, Bayesian optimization
 - Handling imbalanced datasets: SMOTE, weighted loss functions
 - Model compression
- Introduction to ML Deployment
 - Saving and loading models (joblib, pickle)
 - Basic Flask app for serving ML models
- End-to-End ML Pipeline
 - Combining preprocessing, model building, evaluation, and deployment using scikit-learn pipelines

Project:

1. Regression Project: Predicting house prices using linear and polynomial regression.
2. Classification Project: Customer churn prediction using logistic regression, decision trees, and random forests.

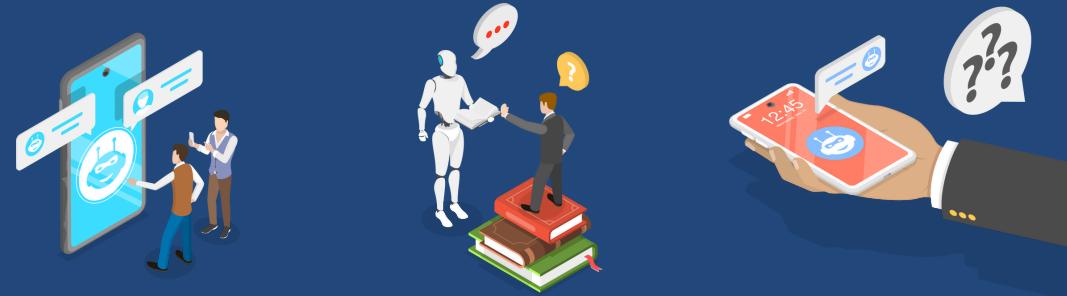


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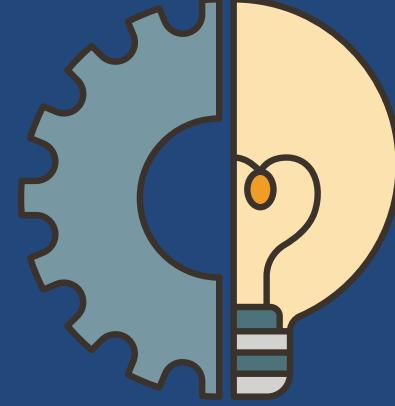


Week 1: Introduction to Deep Learning and Neural Networks

- Introduction to Deep Learning
 - Why deep learning? Real-world applications (image recognition, NLP, autonomous vehicles)
 - Difference between machine learning and deep learning
 - Overview of deep learning frameworks (TensorFlow, PyTorch)
- Neural Networks Basics
 - Biological vs. artificial neurons
 - Architecture: Input, hidden, and output layers
 - Activation functions: ReLU, sigmoid, softmax, tanh
 - Cost functions: MSE, cross-entropy
- Feedforward Neural Networks
 - Forward and backward propagation, gradient descent, learning rate tuning
 - Hands-on: Building a simple neural network from scratch in TensorFlow/Keras

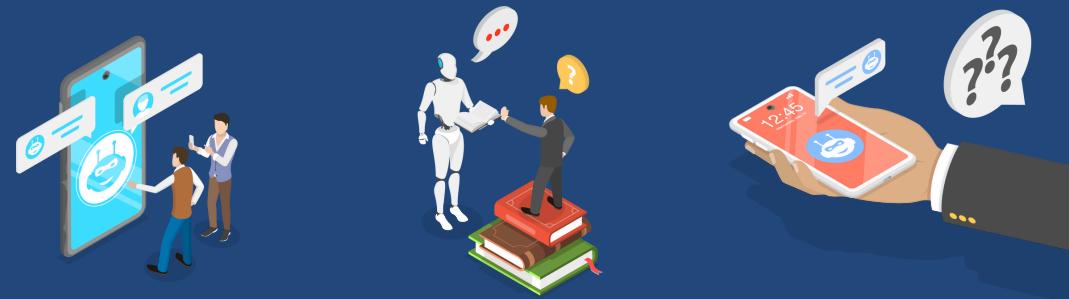


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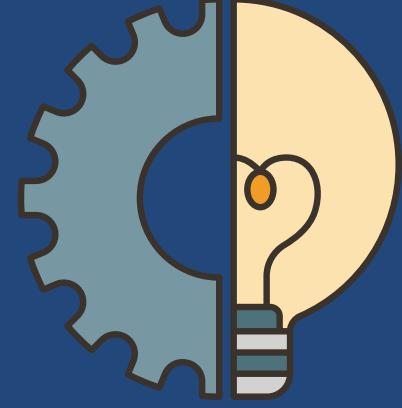


Week 2: Convolutional Neural Networks (Image-Based Project)

- Introduction to Convolutional Neural Networks (CNNs)
 - Convolution operation, filters/kernels, feature maps, auto-encoding
 - Pooling layers (max pooling, average pooling), flattening, fully connected layers
- Architectures and Use Cases
 - Famous architectures: VGG, ResNet, EfficientNet (brief overview)
 - Common use cases: Object detection, image classification
- Project (Part 1): Image Classification
 - Dataset: CIFAR-10 or Fashion-MNIST
 - Preprocessing: Resizing, normalizing, data augmentation (rotation, flipping)
 - Building a CNN model using TensorFlow/Keras
 - Training, validation, and testing the model

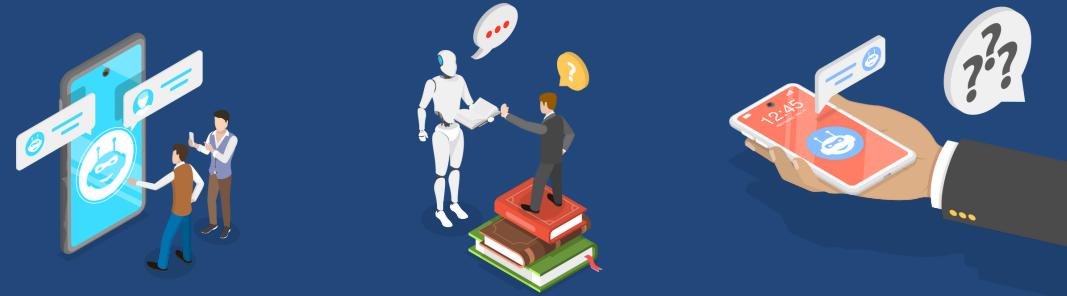


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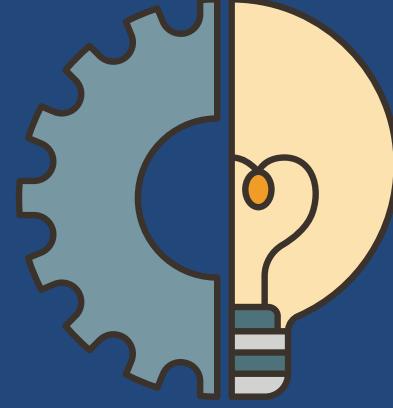


Week 3: Optimization and Transfer Learning (Image-Based Project Continuation)

- Optimization Techniques
 - Optimizers: SGD, Adam, RMSProp
 - Regularization: Dropout, L2 regularization
 - Learning rate schedules: Step decay, exponential decay
- Transfer Learning
 - Concept and benefits of transfer learning
 - Using pre-trained models (e.g., ResNet, MobileNet)
 - Fine-tuning and feature extraction
- Project (Part 2): Improving Image Classification Model
 - Apply transfer learning to improve accuracy
 - Fine-tune pre-trained model (e.g., MobileNet)
 - Evaluate and interpret the model's performance

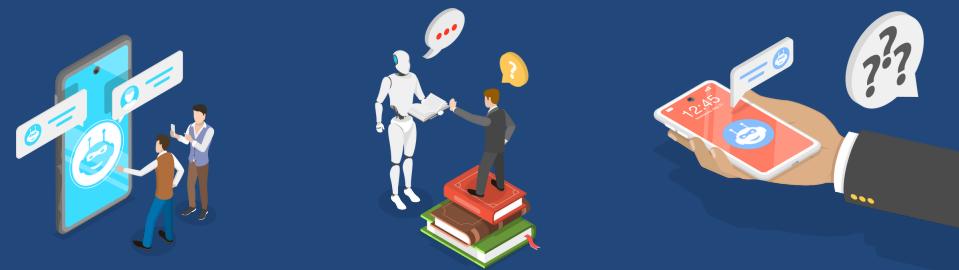


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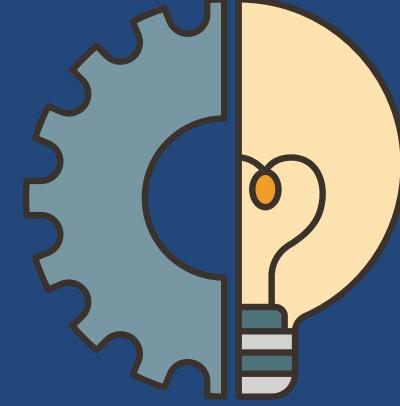


Week 4: Recurrent Neural Networks (RNNs) and Natural Language Processing (NLP-Based Project)

- Introduction to RNNs
 - Sequential data, limitations of feedforward networks
 - RNN structure, vanishing gradient problem
- Advanced RNNs: LSTMs and GRUs
 - Long Short-Term Memory (LSTM): Gates (forget, input, output), cell state
 - Gated Recurrent Units (GRUs): Simplified LSTM architecture
- Basics of Natural Language Processing
 - Tokenization, stemming, lemmatization, word embeddings (Word2Vec, GloVe)
 - Text preprocessing: Removing stopwords, handling punctuation, lowercasing
- Project (Part 1): Sentiment Analysis on Text Data
 - Dataset: IMDb or Twitter Sentiment Dataset
 - Preprocessing: Tokenization, padding sequences
 - Building an RNN/LSTM model for sentiment classification



DEEP LEARNING



Week 5: Transformers and Advanced NLP Techniques (NLP-Based Project Continuation)

- Introduction to Transformers
 - Concept of attention, encoder-decoder architecture, multi-head attention
 - Overview of BERT, GPT, and their applications in NLP
- Fine-Tuning Pre-Trained NLP Models
 - Hugging Face library: Loading and fine-tuning BERT or GPT for text classification
 - Handling large datasets and transfer learning for NLP
- Project (Part 2): Improving NLP Sentiment Analysis
 - Fine-tune BERT/GPT on the same sentiment dataset
 - Compare performance with RNN/LSTM model
 - Model evaluation: Precision, recall, F1-score, confusion matrix

Key Deliverables

1. Project 1 (Image-Based): Image classification model with CNNs, enhanced using transfer learning.
2. Project 2 (NLP-Based): Sentiment analysis model with LSTMs, improved using a fine-tuned transformer model like BERT.

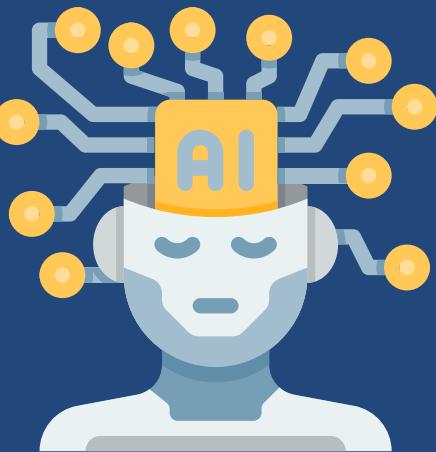


GENERATIVE AI



Week 1: Foundations of Generative AI and GANs

- Introduction to Generative AI
 - What is Generative AI? Applications: Image generation, text generation, deepfakes, AI art
 - Difference between discriminative and generative models
 - Generative AI in industries: Gaming, healthcare, content creation
- Introduction to Generative Adversarial Networks (GANs)
 - Architecture: Generator and discriminator
 - Adversarial training, loss functions, mode collapse
- Hands-on Project (Part 1): Building a Basic GAN
 - Dataset: MNIST (handwritten digits)
 - Preprocessing: Data normalization
 - Implementing a simple GAN using PyTorch or TensorFlow
 - Training the GAN and generating basic images

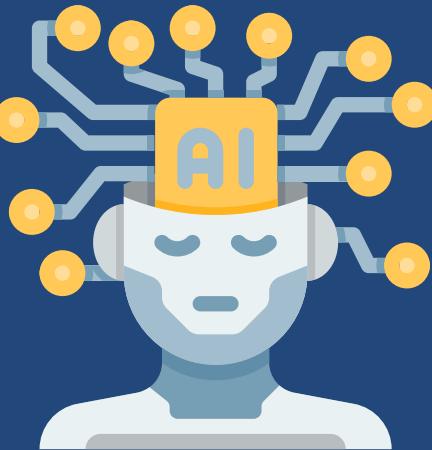


GENERATIVE AI



Week 2: Advanced GANs and Variational Autoencoders (VAEs)

- Advanced GAN Concepts
 - Conditional GANs (cGANs): Generating specific categories (e.g., numbers, colors)
 - CycleGANs: Style transfer for unpaired image datasets (e.g., turning photos into paintings)
 - Progressive GANs and StyleGANs: High-resolution image generation
- Introduction to Variational Autoencoders (VAEs)
 - Difference between autoencoders and VAEs
 - Understanding the encoder-decoder structure and latent space
 - Applications of VAEs: Image reconstruction, anomaly detection
- Hands-on Project (Part 2): Image-to-Image Translation
 - Dataset: Edges-to-shoes or similar paired datasets
 - Implementing a CycleGAN for style transfer or image-to-image mapping

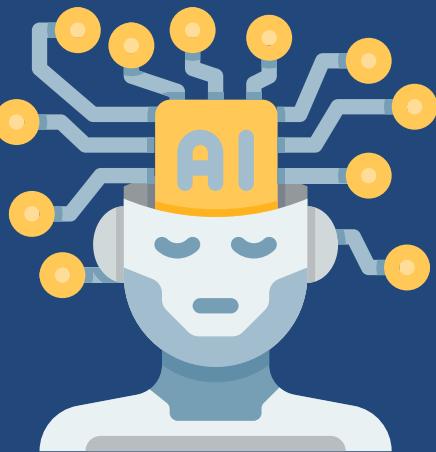


GENERATIVE AI



Week 3: Generative AI for Text and Language Models

- Generative Pre-trained Transformers (GPT)
 - Architecture: Transformers, self-attention mechanism, positional encoding
 - Overview of GPT models (GPT-2, GPT-3, GPT-4) and their capabilities
- Fine-Tuning Pre-trained Language Models
 - Using Hugging Face library for fine-tuning GPT models
 - Practical applications: Chatbots, text summarization, content generation
- Hands-on Project (Part 3): Text Generation
 - Dataset: News articles or conversational datasets
 - Fine-tuning GPT on a custom dataset for text completion or dialogue generation
 - Evaluating text quality: Perplexity, human evaluation



GENERATIVE AI



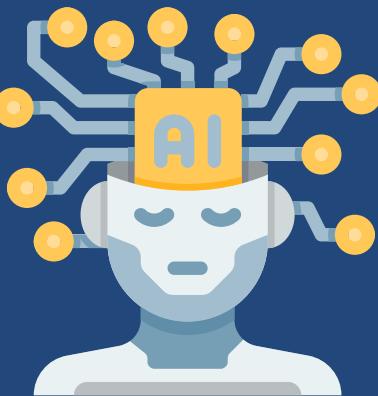
Week 4: Advanced Generative AI Techniques and Applications

- Diffusion Models
 - Understanding diffusion models (e.g., Stable Diffusion, DALL-E)
 - Applications in text-to-image generation and AI art creation
- Ethics and Challenges in Generative AI
 - Bias in generated outputs, intellectual property issues, misuse of generative models (e.g., deepfakes)
 - Best practices and tools for ethical AI development
- Hands-on Project (Part 4): Text-to-Image Generation
 - Tools: Stable Diffusion, OpenAI's DALL-E API
 - Generating images based on textual descriptions
 - Fine-tuning image generation models on custom datasets for specific styles

Key Deliverables

1. Project 1: Basic GAN for image generation (e.g., MNIST handwritten digits).
2. Project 2: CycleGAN for style transfer or image-to-image translation.
3. Project 3: Fine-tuned GPT for text generation (e.g., creating human-like dialogues or articles).

Project 4: Text-to-image generation using Stable Diffusion or DALL-E.

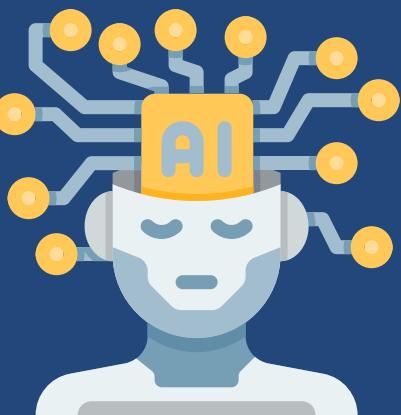


BUILDING AI AGENT



Week 1: Introduction to AI Agents and LangChain Basics

- What Are AI Agents?
 - Definition and purpose of AI agents
 - Applications in customer service, task automation, and personalized recommendations
 - Key components: Perception, decision-making, and action
- Introduction to LangChain
 - What is LangChain? Importance of modular frameworks for building LLM-powered applications
 - Core concepts:
 - Chains: Sequential and parallel workflows
 - Agents: Decision-making systems powered by LLMs
 - Tools: Integrating APIs, web scraping, and database queries
- Hands-on Project (Part 1): Task Automation with LangChain
 - Use case: Automating data retrieval and summarization from multiple online sources
 - Implementing:
 - Chain for querying APIs and processing results
 - LLM-based summarization with OpenAI models
 - Tool integration (e.g., Google Search API)

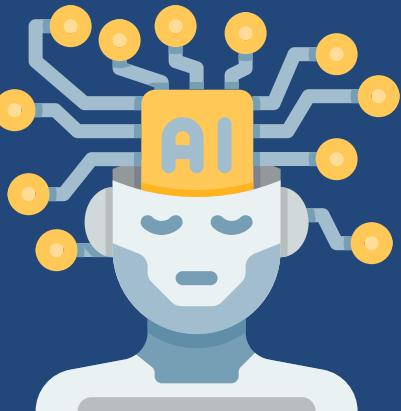


BUILDING AI AGENT



Week 2: Advanced AI Agent Concepts and CrewAI Project

- Advanced AI Agent Design
 - Memory in AI agents: Types (short-term, long-term), vector stores, and persistence with LangChain
 - Decision-making: Action planning using tools like Python functions and structured APIs
 - Error handling: Managing unexpected outputs or failures gracefully
- Introduction to CrewAI
 - Overview of CrewAI: A simplified tool for building AI agents without deep coding knowledge
 - How CrewAI complements LangChain for rapid prototyping and iterative development



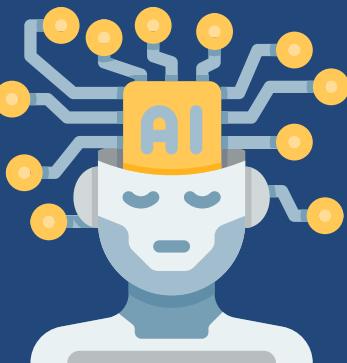
BUILDING AI AGENT



- Hands-on Project (Part 2): Task-Oriented AI Agent with CrewAI and LangChain
 - Use case: Building an AI assistant for project management tasks
 - Features to implement:
 - Querying deadlines and project updates from a spreadsheet
 - Summarizing team performance using pre-defined metrics
 - Suggesting next steps or optimizations for workflows
 - Tools and integrations:
 - Google Sheets API for task data
 - OpenAI or Hugging Face LLMs for summarization and recommendations
 - Memory modules to maintain conversation context
 -

Key Deliverables

1. Project 1 (LangChain): AI task automation agent for real-world data retrieval and summarization.
2. Project 2 (CrewAI + LangChain): Task-oriented AI agent for project management, combining multiple tools and memory for enhanced decision-making.

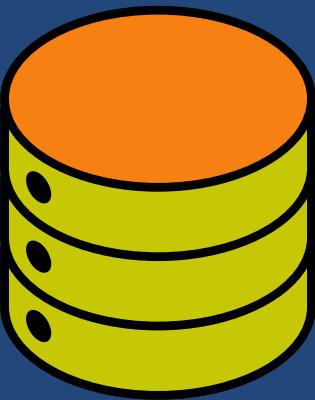


SQL



Week 1: Introduction to SQL and Database Design

- **Understanding Databases**
 - Relational databases vs. non-relational databases
 - Database management systems (DBMS): MySQL, PostgreSQL, SQLite
 - Introduction to tables, rows, columns, and relationships
- **Basic SQL Queries**
 - SELECT statement, filtering with WHERE clause
 - Sorting results with ORDER BY
 - Aggregating data: COUNT, SUM, AVG, MAX, MIN
- **Basic Joins**
 - Inner join, left join, right join, full outer join
 - Join conditions and understanding relationships
- **Hands-on Project (Part 1): Data Extraction from a Sample Dataset**
 - Dataset: Sales data, customer data, or ecommerce data
 - Extracting relevant data using SELECT and filtering with WHERE
 - Sorting results, performing basic aggregation (e.g., total sales, customer counts)

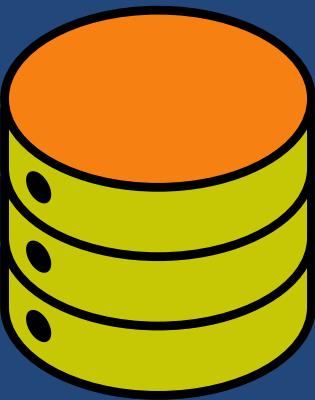


SQL

SQL

Week 2: Advanced SQL Queries and Subqueries

- **Advanced Joins**
 - Self-joins, cross joins
 - Combining multiple tables with complex join conditions
- **Subqueries**
 - Subqueries in SELECT, WHERE, and FROM clauses
 - Using subqueries for filtering and calculating derived metrics
- **Grouping Data**
 - GROUP BY, HAVING clause for advanced aggregation
 - Using aggregate functions in grouped data (e.g., total sales by region, average order size by product)
- **Hands-on Project (Part 2): Building Reports with Advanced Queries**
 - Dataset: Sales and customer transactions
 - Building reports with GROUP BY to analyze total sales by region, average transaction value
 - Using subqueries to identify top-performing products, customers, or regions

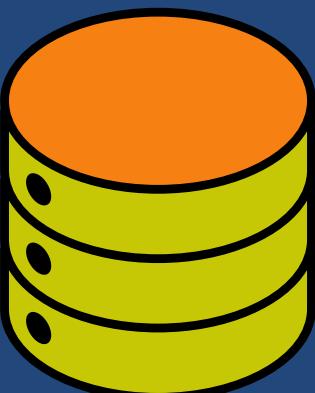


SQL

SQL

Week 3: Data Modification and Indexing

- **Data Insertion, Updates, and Deletions**
 - INSERT INTO statement for adding records
 - UPDATE statement for modifying data, handling data integrity
 - DELETE statement and handling referential integrity (e.g., cascading deletes)
- **Transactions and ACID Properties**
 - Understanding transactions (COMMIT, ROLLBACK)
 - Ensuring data consistency and atomicity
- **Indexes and Query Optimization**
 - Importance of indexing for fast query performance
 - Creating and using indexes
 - Query execution plans and optimizing complex queries
- **Hands-on Project (Part 3): Data Management and Optimization**
 - Dataset: Employee records and project assignments
 - Inserting, updating, and deleting records while maintaining integrity
 - Optimizing queries using indexes, improving performance on large datasets
 - Analyzing query performance using execution plans



SQL



Week 4: Complex SQL Queries, Views, and Company-Specific Problems

- **Complex Queries**

- Window functions (e.g., ROW_NUMBER(), RANK(), LEAD(), LAG())
- Working with dates and time intervals (DATE, TIMESTAMP, DATE_DIFF, DATE_ADD)
- String manipulation and pattern matching (LIKE, REGEXP)

- **Views and Stored Procedures**

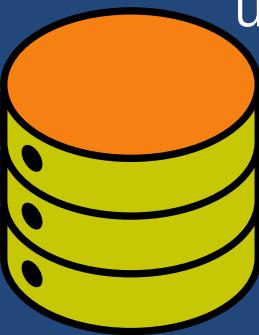
- Creating and using views for reusable queries
- Introduction to stored procedures for automating SQL logic
- Triggers and events for automated actions based on data changes

- **Company-Specific Problem Solving**

- Solving real-world company-specific SQL problems based on the industry (e.g., e-commerce, finance, healthcare)
 - Example: Analyzing customer churn in an e-commerce company
 - Example: Financial report generation for a company
 - Example: Analyzing employee performance and project completion in a corporate environment

- **Hands-on Project (Part 4): Company-Specific SQL Project**

- Use case: Building complex reports and automating data retrieval for business operations
- Working with company datasets to solve industry-specific problems using SQL queries, joins, and optimizations



SQL



Week 4: Complex SQL Queries, Views, and Company-Specific Problems

- **Complex Queries**

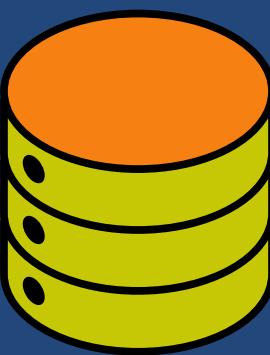
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SQL

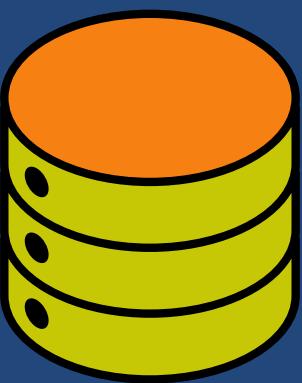


- **Hands-on Project (Part 4): Company-Specific SQL Project**

- Use case: Building complex reports and automating data retrieval for business operations
- Working with company datasets to solve industry-specific problems using SQL queries, joins, and optimizations

Key Deliverables

1. **Project 1:** Basic SQL report building with dataset extraction, filtering, and aggregation.
2. **Project 2:** Advanced report generation with multiple joins, subqueries, and groupings.
3. **Project 3:** Data management, transaction handling, and query optimization using indexing and execution plans.
4. **Project 4:** Solving a company-specific SQL problem (e.g., customer churn analysis, financial report generation).



CAPSTONE PROJECTS



Capstone Projects (+2 Projects)

💡 End-to-End ML Pipeline

- Data Preprocessing: Handling missing data, normalization, encoding categorical variables, and exploratory data analysis (EDA)
- Feature Engineering: Creating meaningful features, dimensionality reduction, and feature selection techniques
- Model Building: Training multiple models (e.g., regression, classification, ensemble methods) and hyperparameter tuning for optimization
- Evaluation: Assessing model performance with metrics like accuracy, precision, recall, F1-score, ROC-AUC, and confusion matrix
- Deployment: Building APIs, containerization with Docker, and integrating models into live systems

Showcase your expertise by solving real-world problems from start to finish!





ADDITIONAL PERKS

- Resume Building Guidance
- 5 Mock Interviews
- Lifetime Discord Community Access For QA
- Certificate of Completion
- 1 Month Unpaid Internship on Real World Project



WHY US ?



- Up-to-Date and Beginner-Friendly Course
- Hands-On Industry Projects
- Proven Track Record
- Fully Live Learning
- Experienced Mentors
- Affordable and Value-Packed
- Flexible for Everyone
- Support and Mentorship

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