

# ALVA'S INSTITUTE OF ENGINEERING AND TECHNOLOGY

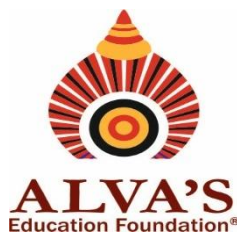
A Unit of Alva's Education Foundation (R)

(Affiliated to Visvesvaraya Technological University, Belagavi.

Approved by AICTE, New Delhi & & Recognized by Government of Karnataka)

Shobhavana Campus, Mijar, Moodbidri- 574 225, Mangalore, D.K., Karnataka

State.



## YEARLY REPORT



ON

## PROJECTS AND LEARNING

*Report submitted in fulfillment for the Completion of*

*Projects*

for

**EDWIN'S LAB**

By

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**4AL23IS024**

**ACADEMIC YEAR 2024-25**

**Submitted to**

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Artificial Intelligence (AI) is the science of building machines capable of performing tasks that normally require human intelligence, such as decision-making, recognition, and problem-solving.

Machine Learning (ML) is a subset of AI where machines learn patterns from data without explicit programming.

Deep Learning (DL) is a subset of ML that uses multi-layered neural networks to automatically extract complex features and make predictions.

**Real-life examples:**

- AI: Chatbots, self-driving cars
  - ML: Spam email detection, stock price prediction
  - DL: Face recognition, language translation
- 

## Machine Learning (ML)

Machine Learning is the process of teaching computers to learn patterns from data.

**Types of ML:**

- Supervised Learning: Models trained with labeled data (input → output)
  - Unsupervised Learning: Models trained with unlabeled data to find patterns or clusters
  - Reinforcement Learning: Agents learn by trial-and-error using rewards/punishments
- 

### Supervised Learning Algorithms

#### Linear Regression

- Predicts continuous values
- Equation:  $y = \beta_0 + \beta_1 x + \epsilon$
- Cost function (MSE):  $J(\theta) = (1/2m) \sum (h_{\theta}(x_i) - y_i)^2$
- Applications: Predicting student scores, sales forecasting

## Project(liner regression)

The project applies **linear regression** to predict house prices based on features such as area, number of bedrooms, location, and other property attributes. The dataset is preprocessed (handling missing values, encoding categorical data, and scaling). The model learns the

relationship between input features and price by fitting a best-fit line. Performance is evaluated using metrics like **R<sup>2</sup> score, Mean Squared Error (MSE), and Root Mean Squared Error (RMSE)**. Results show that linear regression provides a simple and interpretable baseline for price prediction, though more complex models (e.g., Random Forest or XGBoost) can improve accuracy.

## Logistic Regression

🔍 **Purpose:** Used for binary and multiclass classification tasks.

🔍 **Hypothesis function:**

- Uses Sigmoid function  $\rightarrow \sigma(z) = 1 / (1 + e^{(-z)})$ , which maps values into (0,1).

🔍 **Decision boundary:**

- If  $P(y=1|x) \geq 0.5 \rightarrow$  class 1, else class 0.

🔍 **Cost function (Log Loss):**

- $J(\theta) = -(1/m) \sum [ y_i \log(h\theta(x_i)) + (1-y_i) \log(1-h\theta(x_i)) ]$

🔍 **Gradient Descent:**

- Used to optimize parameters  $\theta$  to minimize cost.

🔍 **Regularization (L1/L2):**

- Prevents overfitting by penalizing large weights.

🔍 **Types:**

- Binary Logistic Regression  $\rightarrow$  two classes (e.g., spam vs. not spam).
- Multinomial Logistic Regression (Softmax Regression)  $\rightarrow$  more than two classes.

🔍 **Evaluation Metrics:**

- Accuracy, Precision, Recall, F1-score, ROC-AUC.

🔍 **Applications:**

- Email spam detection, medical diagnosis, credit scoring, fraud detection, sentiment analysis.

## Project

### Wine quality prediction

The project uses **logistic regression** to predict wine quality (good or bad) based on chemical features. It shows which factors (like alcohol, acidity, and sulphates) most influence quality, achieving around **70–80% accuracy** as a baseline model.

The screenshot shows a Jupyter Notebook in VS Code with the following code and output:

```

34, # total sulfur dioxide
0.9978, # density
3.51, # pH
0.56, # sulphates
9.4]]) # alcohol
log_model_pred=logmodel.predict(sample)
r_forest_pred=r_forest.predict(sample)
knn_model_pred=knn_model.predict(sample)
xg_boost_model_pred=xgboost_model.predict(sample)
xg_boost_model_pred=[ if pred>0.5 else 0 for pred in xg_boost_model_pred]

[142]

print("prediction of the logistic model",log_model_pred)
print("prediction of Randomforestclassifier model",r_forest_pred)
print("prediction of the KNeighborsClassifier",knn_model_pred)
print("prediction of the XGBoost Classifier",xg_boost_model_pred)

[143]

prediction of the logistic model [0]
prediction of Randomforestclassifier model [0]
prediction of the KNeighborsClassifier [0]
prediction of the XGBoost Classifier [0]

[144]

from collections import Counter
predictions=[log_model_pred,r_forest_pred,knn_model_pred,xg_boost_model_pred]
flat_preds = np.array(predictions).flatten()
output=Counter(flat_preds).most_common(1)[0][0]
print("final prediction",output)

[145]

Final prediction 0

[1]

if output==1:

```

The output shows the prediction of the logistic model, Random Forest classifier, KNeighborsClassifier, and XGBoost classifier. The final prediction is 0.

# Decision Tree

- Tree-like structure for classification/regression
- Gini Impurity:  $\text{Gini}(p) = 1 - \sum p_i^2$
- Entropy:  $H(S) = - \sum p_i \log_2(p_i)$
- Applications: Customer segmentation, loan approval
- **[Insert diagram/screenshot here]**

## K-Nearest Neighbors (KNN)

- Classifies based on nearest neighbors
- Euclidean distance:  $d(p,q) = \sqrt{\sum (p_i - q_i)^2}$

- Applications: Movie recommendations, image classification
  - **[Insert diagram/screenshot here]**
- 

## Unsupervised Learning Algorithms

### Clustering (K-Means)

- Groups similar data points
- Applications: Market segmentation, customer behavior analysis
- **[Insert diagram/screenshot here]**

### Dimensionality Reduction (PCA)

- Reduces dataset features while retaining important information
  - Applications: Image compression, feature visualization
  - **[Insert diagram/screenshot here]**
- 

## Reinforcement Learning

- Agents learn to act in an environment to maximize cumulative reward
  - Examples: Self-driving cars, game-playing AI
  - **[Insert diagram/screenshot here]**
- 

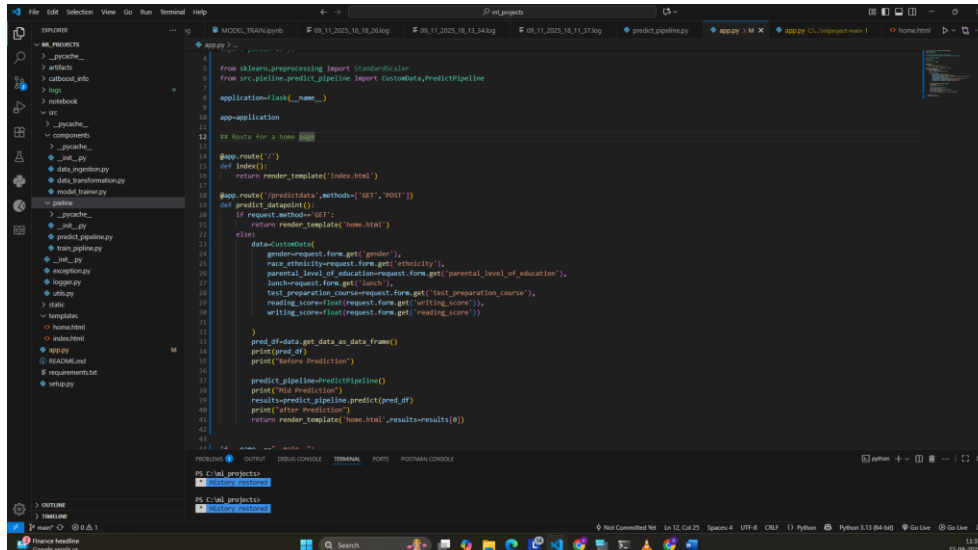
## Student Score Prediction Project (End-to-End ML Project)

- Overview: Predict student Maths score using ML regression models based on features like gender, parental education, writing score, and reading score
- Workflow: Data preprocessing → Model training → Flask web app deployment
- Tech Stack: Python, Pandas, Sklearn, Flask, HTML/CSS
- **[Insert UI screenshot here]**
- **[Insert VS Code project structure screenshot here]**
- **[Insert results screenshot here]**

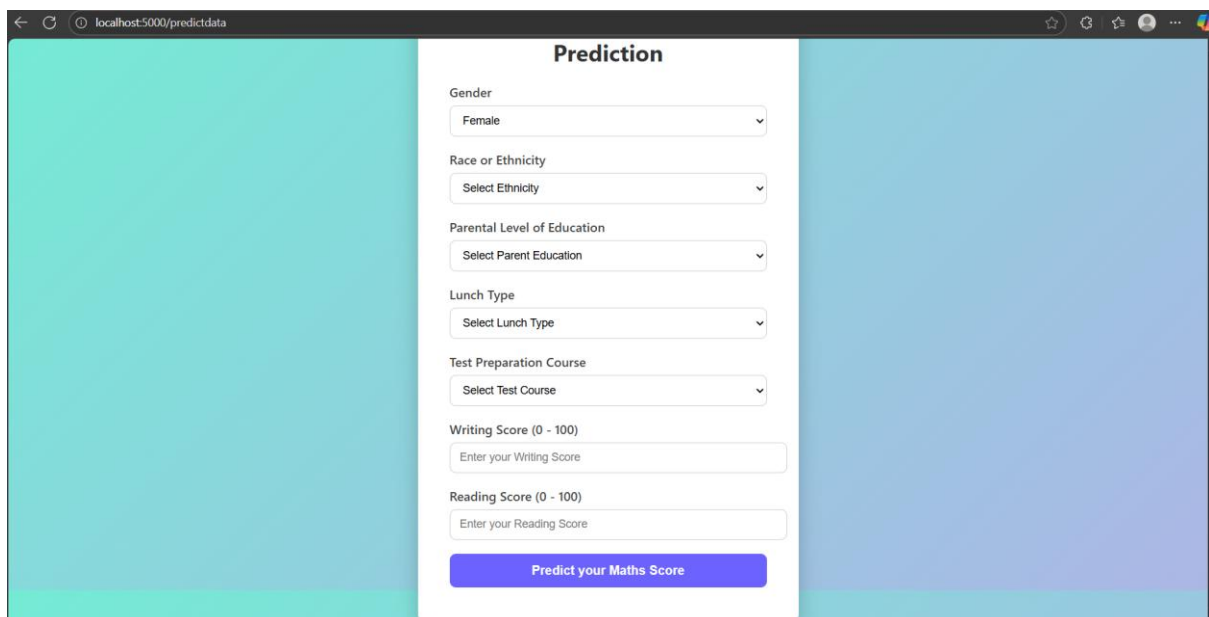
## Summary

This project is an **end-to-end ML pipeline** with **data preprocessing, model training, and prediction**, deployed via a **Flask web app**. It's modular, with separate components for ingestion, transformation, training, and pipelines, plus logging, exception handling, and a simple HTML interface for user input and results.

## Sample



```
1 from sklearn.preprocessing import StandardScaler
2 from src.pipeline_predict_pipeline import customData, PredictPipeline
3
4 application=flask(__name__)
5
6 app=application
7
8 # Route for a home page
9
10 @app.route('/')
11 def index():
12     return render_template("index.html")
13
14 @app.route('/predictdata', methods=['GET', 'POST'])
15 def predict_datapoint():
16     if request.method=='GET':
17         return render_template("home.html")
18     else:
19         data=request.form.get('gender'),
20         race=request.form.get('ethnicity'),
21         parental_level_of_education=request.form.get('parental_level_of_education'),
22         lunch=request.form.get('lunch'),
23         test_preparation_course=request.form.get('test_preparation_course'),
24         reading_score=float(request.form.get('reading_score')),
25         writing_score=float(request.form.get('writing_score'))
26
27     pred_df=data.get_data_as_data_frame()
28     print(pred_df)
29     print("Before Prediction")
30
31     predict_pipeline=PredictPipeline()
32     print("MLA Prediction")
33     results=predict_pipeline.predict(pred_df)
34     print("after Prediction")
35     return render_template("home.html", results=results[0])
36
37 if __name__ == '__main__':
38     app.run(debug=True)
```



The web application interface is titled "Prediction" and is located at the URL `localhost:5000/predictdata`. It features a series of dropdown menus and text input fields for user input, followed by a "Predict your Maths Score" button.

**Prediction**

Gender:

Race or Ethnicity:

Parental Level of Education:

Lunch Type:

Test Preparation Course:

Writing Score (0 - 100):

Reading Score (0 - 100):

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## Deep Learning (DL)

Deep Learning uses multi-layered neural networks to learn complex patterns automatically.

### Artificial Neural Networks (ANN)

- Structure: Input → Hidden Layers → Output
- Forward pass:  $a(l) = f(W(l)a(l-1) + b(l))$
- Backpropagation updates weights to minimize error
- Applications: Classification, regression
- **[Insert diagram/screenshot here]**

## Projects

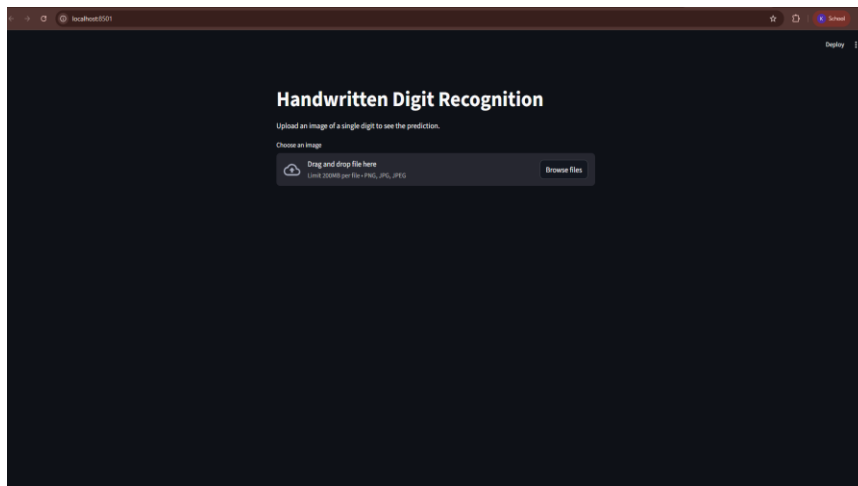
### handwritten-digit-recognition-cnn

The project uses a **Convolutional Neural Network (CNN)** to recognize handwritten digits from the **MNIST dataset (0–9)**. The data is preprocessed (normalization, reshaping, and one-hot encoding), then passed through convolutional, pooling, and fully connected layers to learn visual patterns. The model is trained to classify digits with high accuracy (often above **98%**). Finally, it can predict digits from user-drawn images or test samples, showcasing how deep learning can be applied to **image classification tasks**.

## Samples

```

1 import tensorflow as tf
2 import numpy as np
3 from PIL import Image, ImageOps
4 import streamlit as st
5 import os
6
7 # Load the model from the correct path
8 # The '..' navigates up one directory to find the 'model' folder
9 try:
10     from tensorflow.keras.models import load_model
11 except Exception as e:
12     st.error(f"Error loading the model. Make sure 'mnist_model.h5' is in the 'model' directory at the root level of your project. Error: {e}")
13     st.stop()
14
15 model = load_model("mnist_model.h5")
16
17 def predictImage(image):
18     """
19     Preprocesses the input image and makes a prediction using the loaded model.
20     """
21     # Convert the input image to grayscale
22     image = ImageOps.grayscale(image)
23     # Convert to MNIST size (28x28 pixels)
24     img = image.resize((28, 28))
25     # Normalize pixel values to be between 0 and 1
26     img = np.array(img, dtype="float32") / 255.0
27     # Reshape the image to fit the model's input shape (1, 28, 28, 1)
28     img = img.reshape((1, 28, 28, 1))
29     # Predict the number
30     pred = model.predict(img)
31     return np.argmax(pred[0])
32
33 # Streamlit UI
34 st.title("Handwritten Digit Recognition")
35
36 if __name__ == "__main__":
37     st.write("Upload an image of a single digit to see the prediction.")
38     st.write("Choose an image")
39     st.write("Drag and drop file here")
40     st.write("Browse files")
41
42 if st.button("Predict"):
43     image = st.file_uploader("Upload image")
44     if image:
45         img = image.read()
46         img = Image.open(img)
47         pred = predictImage(img)
48         st.write(f"Predicted digit: {pred}")
49
50 if __name__ == "__main__":
51     st.write("WARNING:tensorflow:From C:\Users\Kiran\AppData\Local\Programs\Python\Python311\Lib\site-packages\keras\src\backend\tensorflow\tf_backend.py:82: The name tf.reset_default_graph is deprecated. Please use tf.compat.v1.reset_default_graph instead.")
52
53 
```



## Convolutional Neural Networks (CNN)

- For image data
- Convolution operation:  $(f * g)(t) = \int f(\tau) g(t-\tau) dt$
- Pooling layers: Max pooling, average pooling
- Applications: Face recognition, object detection
- [Insert diagram/screenshot here]

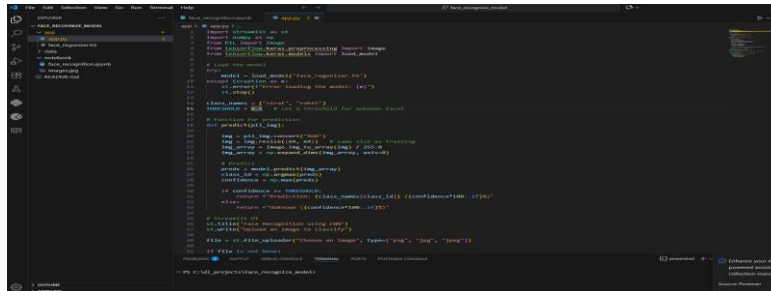
## Project

### CNN Face Classifie

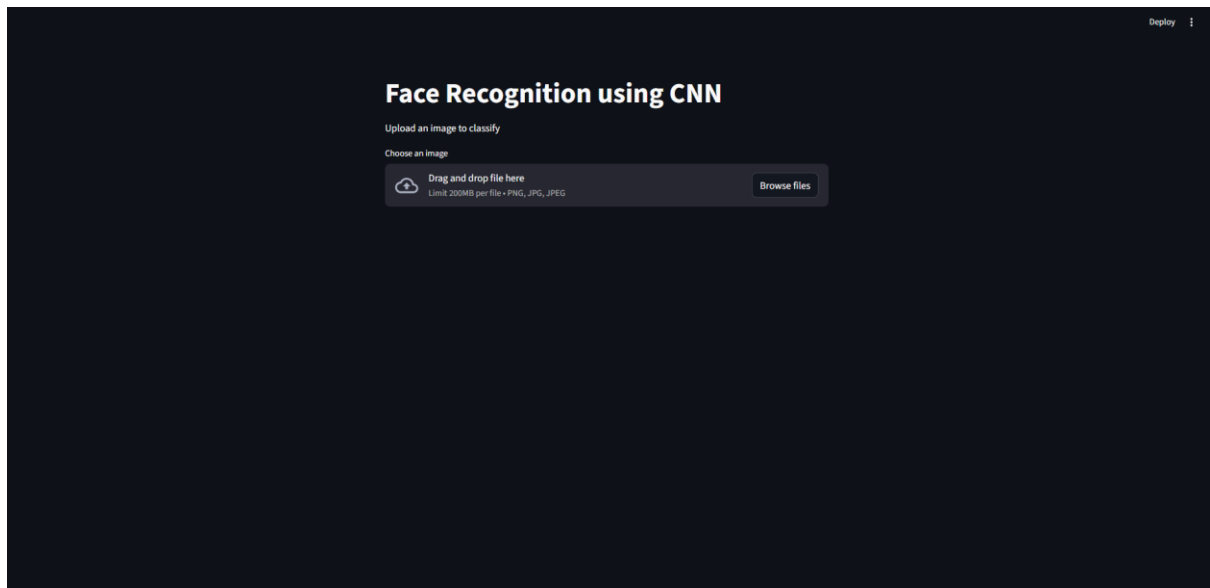


The project uses a Convolutional Neural Network (CNN) to classify faces into categories such as identity, gender, or emotion. The dataset of face images is preprocessed (resizing, normalization, augmentation) and fed into CNN layers that automatically learn facial features. The trained model achieves high accuracy in distinguishing classes and can be deployed for real-time face classification.

## samples



```
1 import os
2 import cv2
3 import numpy as np
4 from keras.preprocessing import image
5 from keras.models import load_model
6
7 # Load the model
8 model = load_model('face_recognition_model.h5')
9
10 # Load the image
11 img = image.load_img('test_image.jpg', target_size=(224, 224))
12 img = image.img_to_array(img)
13 img = np.expand_dims(img, axis=0)
14 img = img.astype('float32') / 255.0
15
16 # Predict the class
17 predictions = model.predict(img)
18
19 # Get the index of the maximum probability
20 predicted_class = np.argmax(predictions)
21
22 # Print the predicted class
23 print('Predicted class:', predicted_class)
```



# Face Recognition Automated Attendance Management System

## Abstract:

This AI-powered Automated Attendance Management System was developed for educational institutions to streamline attendance recording and campus monitoring. The system uses facial recognition technology to automatically mark attendance, prevent proxy entries, and provide secure, tamper-proof attendance logs. It improves efficiency, accuracy, and transparency in academic management.

## Objectives:

- attendance marking using facial recognition.
- Reduce manual workload for teachers and administration.
- Ensure proxy-proof and accurate attendance records.
- Provide instant, secure access to attendance reports for faculty and administration.
- Integrate user-friendly dashboards for students, faculty, and HODs.

## System Design and Methodology:

- **Frontend:** Developed with Next.js and Tailwind CSS for responsive and intuitive UI.
- **Backend:** Built using Go with PostgreSQL database to store attendance, user, and metadata securely.
- **Face Recognition:** Uses RetinaFace for detection and ArcFace for feature extraction and recognition.
- **Workflow:**
  1. Classroom video capture at start/end of classes.
  2. Face detection and alignment.
  3. Anti-spoofing checks to prevent fake attendance.
  4. Embedding generation and matching against the student database.
  5. Attendance marking with timestamp upon successful match.
- **Data Access:** Faculty can instantly fetch processed attendance data via dashboards.

## Key Features:

- Fully automated attendance system.

- Proxy-proof with anti-spoofing measures.
- Real-time student attendance reporting.
- Student dashboards for attendance summary and detailed history.
- Faculty and HOD dashboards with empty-class alerts and inter-branch data sharing.
- Integrated payment management for college fees.
- Botpress chatbot integration for user support.
- Responsive design for web and mobile.

#### **Learning Outcomes:**


- Hands-on experience with AI and ML models for facial recognition.
- Knowledge in full-stack development with Next.js, TypeScript, Go, and PostgreSQL.
- Understanding of real-time data processing, video feeds, and anti-spoofing algorithms.
- Improved skills in designing user-friendly dashboards and workflow automation.

#### **Future Scope:**

- Enhance AI models for faster recognition and higher accuracy.
- Expand certificate verification system to detect fake hospital certificates.
- Implement advanced HOD dashboards for multi-branch event management.
- Integrate more comprehensive student analytics and reporting tools.
-




# CERTIFICATS



3 Courses

- Generative AI: Introduction and Applications
- Generative AI: Prompt Engineering Basics
- Generative AI: Enhance your Data Analytics Career



Sep 12, 2025


**KIRANKUMAR S**


has successfully completed the online, non-credit Specialization


## Generative AI for Data Analysts


Holders of this Professional Certificate have showcased their mastery in generative AI models and tools such as DALL-E, Stable Diffusion, Synthesia, and ChatGPT. They have demonstrated proficiency in various prompt engineering approaches like Interview Pattern, Chain-of-Thought, and Tree-of-Thought. They have also been introduced to commonly used prompt engineering tools like IBM watsonx Prompt Lab, Spellbook, and Dust. In addition, they have learned about the applications of generative AI in data analytics such as testing and development, enhancing data visualizations, filling in missing data points, and ethical considerations. Additionally, the holder successfully passed the final graded exam and quizzes at the end of each lesson.

The online specialization named in this certificate may draw on material from courses taught on-campus, but the included courses are not equivalent to on-campus courses. Participation in this online specialization does not constitute enrollment at this university. This certificate does not confer a University grade, course credit or degree, and it does not verify the identity of the learner.


  
Rav Ahuja  
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IBM Digital Business Group

  
Dr. Pooja  
Instructor and Subject Matter Expert  
Skill-Up Technologies

  
Abhishek Gagneja  
Python and AI Subject Matter Expert

Verify this certificate at:  
<https://coursera.org/verify/specialization/HC14SWTYXF9D>



Sep 10, 2025


**KIRANKUMAR S**

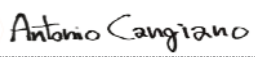
has successfully completed


**Generative AI: Prompt Engineering Basics**

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**COURSE  
CERTIFICATE**



  
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Engineering Manager and AI Specialist  
IBM Digital Business Group

  
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<https://coursera.org/verify/0CA8VRXN17SG>  
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