

Machine Learning Operations

Major Assignment



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GitHub Repository Link – [Github Major Assignment](#)

Phase 1: Repository Setup (using terminal commands)

GitBash Command: mkdir MLOps_Assignment

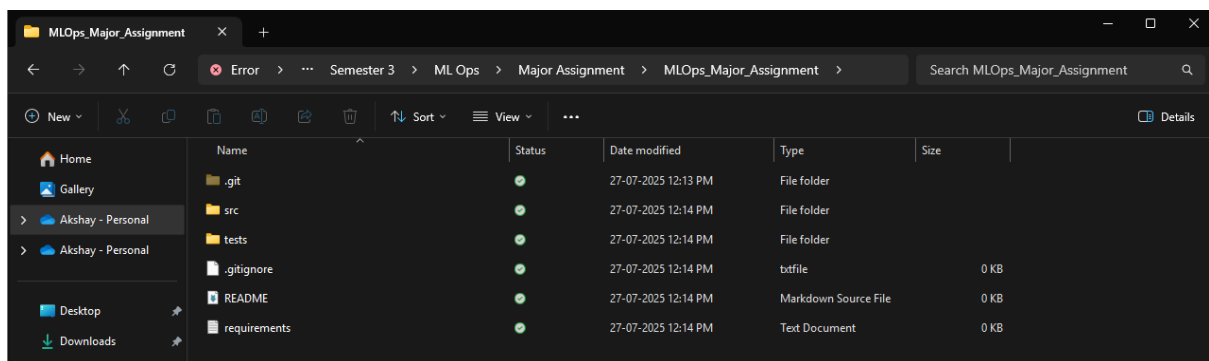
cd MLOps_Assignment

git init

touch README.md .gitignore requirements.txt

mkdir src tests

```
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment
$ mkdir MLOps_Major_Assignment
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment
$ cd MLOps_Major_Assignment
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment
$ git init
Initialized empty Git repository in C:/Users/AKSHAY/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment/.git/
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ touch README.md .gitignore requirements.txt
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ mkdir src tests
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ |
```



Phase 2: Model Training (src/train.py)

Step 1: Create src/utls.py and add necessary imports

GitBash Command: touch src/utls.py

```
src > src/utls.py > load_data
1 # src/utls.py
2 from sklearn.datasets import fetch_california_housing
3 from sklearn.model_selection import train_test_split
4
5 def load_data():
6     """Loads the California Housing dataset and splits it into training and testing sets."""
7     housing = fetch_california_housing()
8     X, y = housing.data, housing.target
9     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
10    return X_train, X_test, y_train, y_test
```

Step 2: Create src/train.py and add the training logic

GitBash Command: touch src/train.py

```
train.py U X
src > train.py > ...
1  import joblib
2  from sklearn.linear_model import LinearRegression
3  from sklearn.metrics import r2_score, mean_squared_error
4  from src.utils import load_data
5  import os
6
7  def train_model():
8      print("Loading data...")
9      X_train, X_test, y_train, y_test = load_data()
10
11     print("Training Linear Regression model...")
12     model = LinearRegression()
13     model.fit(X_train, y_train)
14
15     print("Evaluating model...")
16     y_pred = model.predict(X_test)
17     r2 = r2_score(y_test, y_pred)
18     mse = mean_squared_error(y_test, y_pred)
19
20     print(f"R^2 Score: {r2:.4f}")
21     print(f"Mean Squared Error (Loss): {mse:.4f}")
22
23     os.makedirs('models', exist_ok=True)
24     model_path = os.path.join('models', 'linear_regression_model.joblib')
25     joblib.dump(model, model_path)
26     print(f"Model saved to {model_path}")
27
28 if __name__ == "__main__":
29     train_model()
```

Step 3: Update requirements.txt

```
GNU nano 7.2 requirements.txt Modified
scikit-learn
joblib
```

Step 4: Output of running train.py

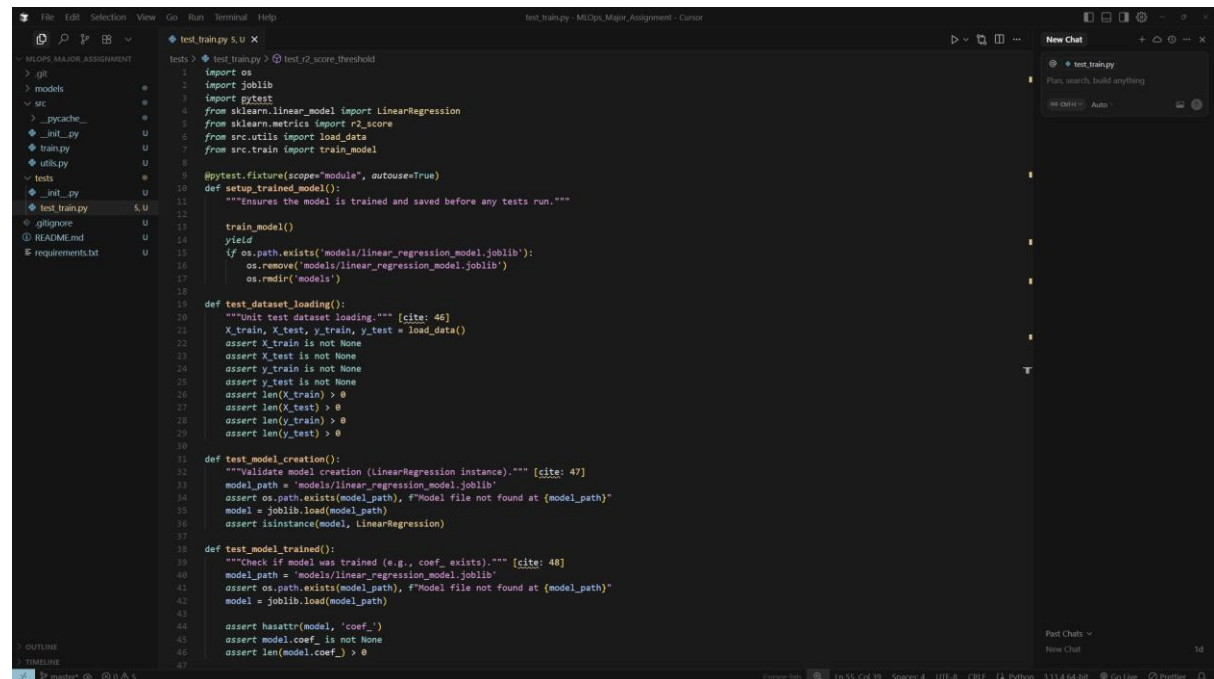
```
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ python -m src.train
Loading data...
Training Linear Regression model...
Evaluating model...
R^2 score: 0.5758
Mean Squared Error (Loss): 0.5559
Model saved to models\linear_regression_model.joblib
```

Phase 3: Testing Pipeline (tests/test_train.py)

Step 1: Create tests/test_train.py

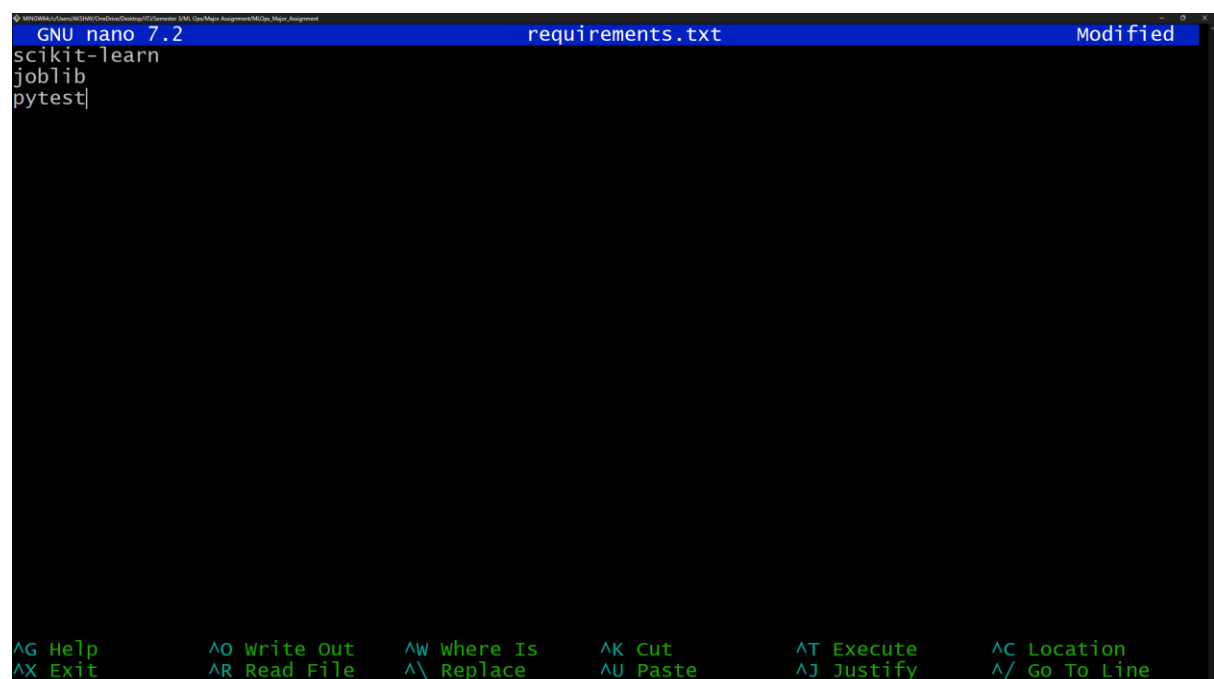
```
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLops_Major_Assignment (master)
$ touch tests/test_train.py
```

Step 2: Add content to tests/test_train.py



```
1 import os
2 import joblib
3 import pytest
4 from sklearn.linear_model import LinearRegression
5 from sklearn.metrics import r2_score
6 from src.utils import load_data
7 from src.train import train_model
8
9 @pytest.fixture(scope="module", autouse=True)
10 def setup_trained_model():
11     """Ensures the model is trained and saved before any tests run."""
12     train_model()
13     yield
14     if os.path.exists('models/linear_regression_model.joblib'):
15         os.remove('models/linear_regression_model.joblib')
16         os.rmdir('models')
17
18 def test_dataset_loading():
19     """Unit test dataset loading.""" [cite: 46]
20     X_train, X_test, y_train, y_test = load_data()
21     assert X_train is not None
22     assert X_test is not None
23     assert y_train is not None
24     assert y_test is not None
25     assert len(X_train) > 0
26     assert len(X_test) > 0
27     assert len(y_train) > 0
28     assert len(y_test) > 0
29
30 def test_model_creation():
31     """Validate model creation (LinearRegression instance).""" [cite: 47]
32     model_path = 'models/linear_regression_model.joblib'
33     assert os.path.exists(model_path), f"Model file not found at {model_path}"
34     model = joblib.load(model_path)
35     assert isinstance(model, LinearRegression)
36
37 def test_model_trained():
38     """Check if model was trained (e.g., coef_exists).""" [cite: 48]
39     model_path = 'models/linear_regression_model.joblib'
40     assert os.path.exists(model_path), f"Model file not found at {model_path}"
41     model = joblib.load(model_path)
42     assert hasattr(model, 'coef_')
43     assert model.coef_ is not None
44     assert len(model.coef_) > 0
```

Step 3: Update requirements.txt (add pytest):



```
GNU nano 7.2 requirements.txt Modified
scikit-learn
joblib
pytest
```

^G Help ^O Write Out ^W Where Is ^K Cut ^T Execute ^C Location
^X Exit ^R Read File ^\ Replace ^U Paste ^J Justify ^_ Go To Line

Step 4: Installing new dependencies:

```
AKSHAY@AKSHAY MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ nano requirements.txt

AKSHAY@AKSHAY MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ pip install -r requirements.txt
Requirement already satisfied: scikit-learn in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from -r requirements.txt (line 1)) (1.3.0)
Requirement already satisfied: joblib in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from -r requirements.txt (line 2)) (1.3.1)
Collecting pytest (from -r requirements.txt (line 3))
  Using cached pytest-8.4.1-py3-none-any.whl.metadata (7.7 kB)
Requirement already satisfied: numpy>=1.17.3 in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from scikit-learn->-r requirements.txt (line 1)) (1.24.3)
Requirement already satisfied: scipy>=1.5.0 in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from scikit-learn->-r requirements.txt (line 1)) (1.11.1)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from scikit-learn->-r requirements.txt (line 1)) (3.2.0)
Requirement already satisfied: colorama>=0.4 in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from pytest->-r requirements.txt (line 3)) (0.4.6)
Collecting iniconfig>=1 (from pytest->-r requirements.txt (line 3))
  Using cached iniconfig-2.1.0-py3-none-any.whl.metadata (2.7 kB)
Requirement already satisfied: packaging>=20 in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from pytest->-r requirements.txt (line 3)) (24.2)
Collecting pluggy<2,>=1.5 (from pytest->-r requirements.txt (line 3))
  Using cached pluggy-1.6.0-py3-none-any.whl.metadata (4.8 kB)
Requirement already satisfied: pygments>=2.7.2 in c:\users\akshay\appdata\local\programs\python\python311\lib\site-packages (from pytest->-r requirements.txt (line 3)) (2.15.1)
Using cached pytest-8.4.1-py3-none-any.whl (365 kB)
Using cached pluggy-1.6.0-py3-none-any.whl (20 kB)
Using cached iniconfig-2.1.0-py3-none-any.whl (6.0 kB)
Installing collected packages: pluggy, iniconfig, pytest
Successfully installed iniconfig-2.1.0 pluggy-1.6.0 pytest-8.4.1

AKSHAY@AKSHAY MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$
```

Step 5: Testing pytest

```
AKSHAY@AKSHAY MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML ops/Major Assignment/MLOps_Major_Assignment (master)
$ pytest
===== test session starts =====
platform win32 -- Python 3.11.4, pytest-8.4.1, pluggy-1.6.0
rootdir: C:\Users\AKSHAY\OneDrive\Desktop\IITJ\Semester 3\ML Ops\Major Assignment\MLOps_Major_Assignment
plugins: anyio-3.7.1, langsmith-0.3.42, typeguard-4.4.4
collected 4 items

tests\test_train.py .... [100%]

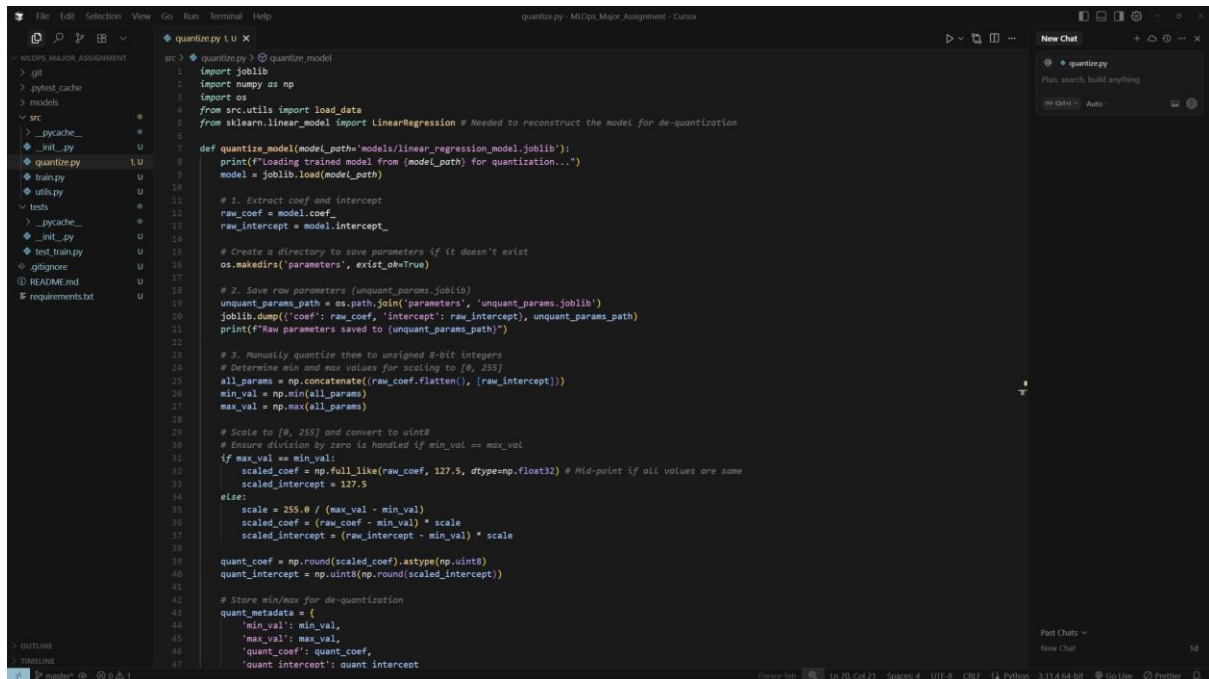
===== 4 passed in 1.04s =====
```

Phase 4: Manual Quantization (src/quantize.py)

Step 1: Create src/quantize.py

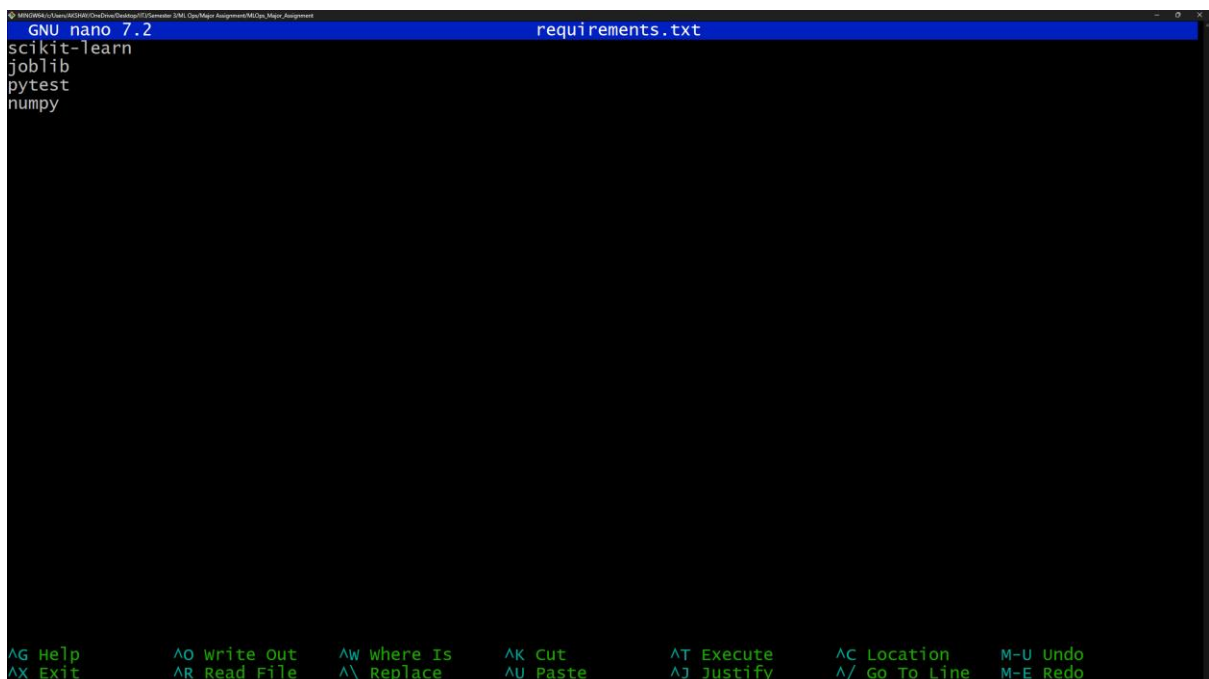
GitBash Command - touch src/quantize.py

Step 2: Add content to src/quantize.py



```
1  import joblib
2  import numpy as np
3  import os
4  from src.utils import load_data
5  from sklearn.linear_model import LinearRegression # Needed to reconstruct the model for de-quantization
6
7  def quantize_model(model_path=models/linear_regression_model.joblib):
8      print(f"Loading trained model from {model_path} for quantization...")
9      model = joblib.load(model_path)
10
11      # 1. Extract coef and Intercept
12      raw_coef = model.coef_
13      raw_intercept = model.intercept_
14
15      # Create a directory to save parameters if it doesn't exist
16      os.makedirs('parameters', exist_ok=True)
17
18      # 2. Save raw parameters (unquant_params.joblib)
19      unquant_params_path = os.path.join('parameters', 'unquant_params.joblib')
20      joblib.dump({'coef': raw_coef, 'intercept': raw_intercept}, unquant_params_path)
21      print(f"Raw parameters saved to {unquant_params_path}")
22
23      # 3. Manually quantize them to unsigned 8-bit integers
24      # Determine min and max values for scaling to [0, 255]
25      all_params = np.concatenate((raw_coef.flatten(), [raw_intercept]))
26      min_val = np.min(all_params)
27      max_val = np.max(all_params)
28
29      # Scale to [0, 255] and convert to uint8
30      # Ensure division by zero is handled if min_val == max_val
31      if max_val == min_val:
32          scaled_coef = np.full_like(raw_coef, 127.5, dtype=np.float32) # Mid-point if all values are same
33          scaled_intercept = 127.5
34      else:
35          scale = 255.0 / (max_val - min_val)
36          scaled_coef = (raw_coef - min_val) * scale
37          scaled_intercept = (raw_intercept - min_val) * scale
38
39      quant_coef = np.round(scaled_coef).astype(np.uint8)
40      quant_intercept = np.uint8(np.round(scaled_intercept))
41
42      # Store min/max for de-quantization
43      quant_metadata = {
44          'min_val': min_val,
45          'max_val': max_val,
46          'quant_coef': quant_coef,
47          'quant_intercept': quant_intercept
48      }
```

Step 3: Update requirements.txt (add numpy)



```
GNU nano 7.2 requirements.txt
scikit-learn
joblib
pytest
numpy
```

AG Help AO Write Out AW Where Is AK Cut AT Execute AC Location M-U Undo
AX Exit AR Read File AL Replace AU Paste AJ Justify A/ Go To Line M-E Redo

Step 4: Install new dependencies:

GitBash Command: pip install -r requirements.txt

Step 5: Run the quantization script

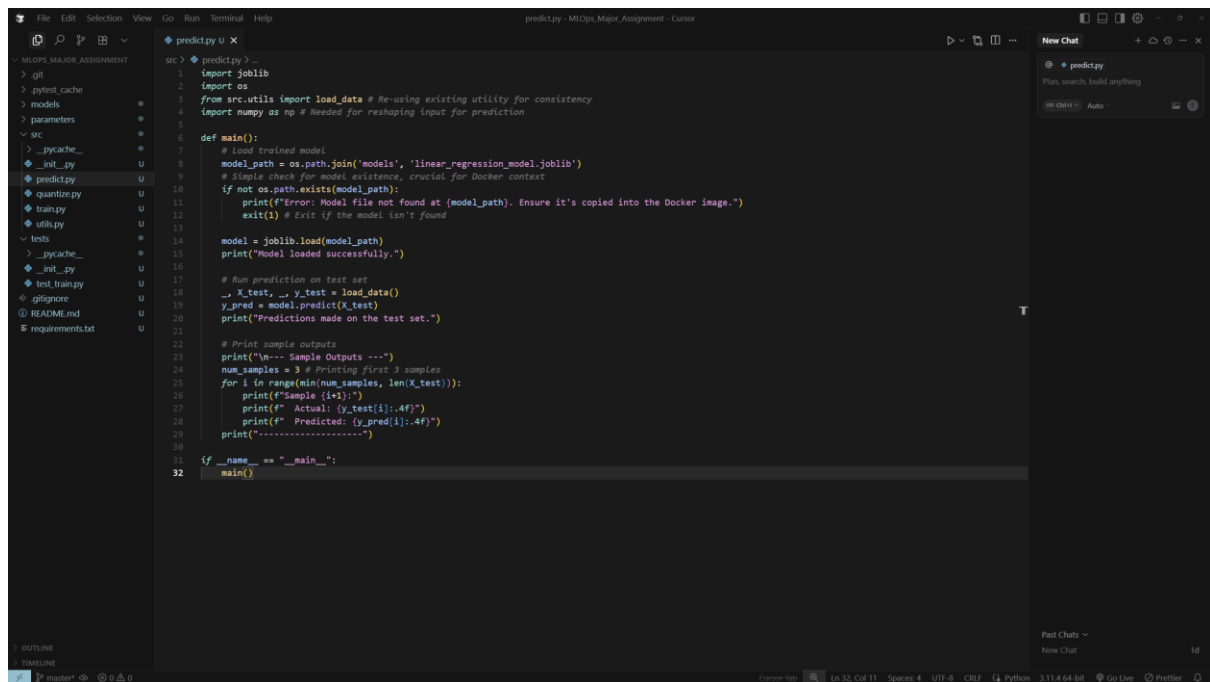
```
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ python -m src.quantize
Loading trained model from models/linear_regression_model.joblib for quantization...
Raw parameters saved to parameters\unquant_params.joblib
Quantized parameters saved to parameters\quant_params.joblib
Performing inference with de-quantized weights...
Sample actual value: 0.4770
Sample original model prediction: 0.7191
Sample de-quantized model prediction: 57.0220
```

Phase 5 : Dockerization - Creating src/predict.py and Dockerfile

Step 1: Create src/predict.py

GitBash Command - touch src/predict.py

Step 2: Add content to src/predict.py

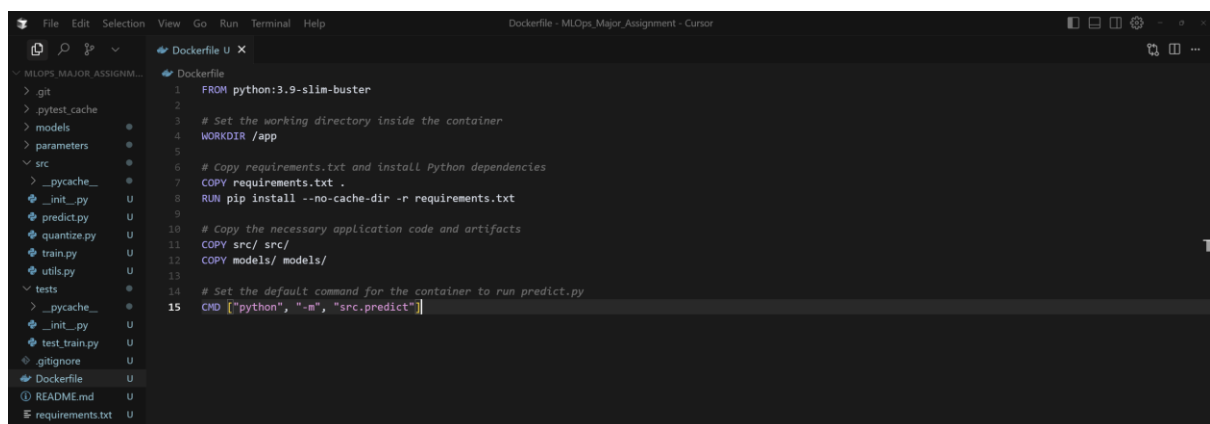


```
1 import joblib
2 import os
3 from src.utils import load_data # Re-using existing utility for consistency
4 import numpy as np # Needed for reshaping input for prediction
5
6 def main():
7     # Load trained model
8     model_path = os.path.join('models', 'linear_regression_model.joblib')
9     # Simple check for model existence, crucial for Docker context
10    if not os.path.exists(model_path):
11        print(f"Error: Model file not found at {model_path}. Ensure it's copied into the Docker image.")
12        exit(1) # Exit if the model isn't found
13
14    model = joblib.load(model_path)
15    print("Model loaded successfully.")
16
17    # Run prediction on test set
18    _, X_test, _, y_test = load_data()
19    y_pred = model.predict(X_test)
20    print("Predictions made on the test set.")
21
22    # Print sample outputs
23    print("\n--- Sample Outputs ---")
24    num_samples = 3 # Printing first 3 samples
25    for i in range(min(num_samples, len(X_test))):
26        print(f"Sample {i+1}:")
27        print(f"    Actual: {y_test[i]:.4f}")
28        print(f"    Predicted: {y_pred[i]:.4f}")
29        print("-----")
30
31 if __name__ == "__main__":
32     main()
```

Step 3: Create Dockerfile

GitBash Command - touch Dockerfile

Step 4: Add content to Dockerfile



```
1 FROM python:3.9-slim-buster
2
3 # Set the working directory inside the container
4 WORKDIR /app
5
6 # Copy requirements.txt and install Python dependencies
7 COPY requirements.txt .
8 RUN pip install --no-cache-dir -r requirements.txt
9
10 # Copy the necessary application code and artifacts
11 COPY src/ src/
12 COPY models/ models/
13
14 # Set the default command for the container to run predict.py
15 CMD ["python", "-m", "src.predict"]
```


Step 5: Running python -m src.train from our project root to ensure models/linear_regression_model.joblib is present.

```
(mlops_env)
AKSHAY@AKSHAY MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ python -m src.train
Loading data...
Training Linear Regression model...
Evaluating model...
R^2 Score: 0.5758
Mean Squared Error (Loss): 0.5559
Model saved to models\linear_regression_model.joblib
```

Step 6: Build the Docker Image

GitBash Command : docker build -t mlops-project:latest .

```
MINGW64~/Users/AKSHAY/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment
AKSHAY@AKSHAY MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (master)
$ docker build -t mlops-project:latest .
#0 building with "desktop-linux" instance using docker driver

#1 [internal] load build definition from Dockerfile
#1 transferring dockerfile: 465B done
#1 DONE 0.0s

#2 [internal] load metadata for docker.io/library/python:3.9-slim-buster
#2 DONE 1.2s

#3 [internal] load .dockerignore
#3 transferring context: 2B done
#3 DONE 0.0s

#4 [internal] load build context
#4 transferring context: 1.44kB done
#4 DONE 0.0s

#5 [1/6] FROM docker.io/library/python:3.9-slim-buster@sha256:320a7a4250aba4249f458872adecf92eea88dc6abd2d76dc5c0f01cac9b5399
0
#5 resolve docker.io/library/python:3.9-slim-buster@sha256:320a7a4250aba4249f458872adecf92eea88dc6abd2d76dc5c0f01cac9b5399 0
.0s done
#5 DONE 0.1s

#5 [1/6] FROM docker.io/library/python:3.9-slim-buster@sha256:320a7a4250aba4249f458872adecf92eea88dc6abd2d76dc5c0f01cac9b5399
0
#5 sha256:2e1c130fa3ec1777a82123374b4c500623959f903c1dd731ee4a83e1f1b38ff2 0B / 3.14MB 0.2s
#5 sha256:824416e234237961c9c5d4f41dfe5b295a3c35a671ee52889bfb08d8e257ec4c 0B / 2.78MB 0.2s
#5 sha256:84c8c79126f669beec1dcf6f34cd88094471745570c19c29b465dfa7db1fdabd 0B / 243B 0.2s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 0B / 11.04MB 0.2s
#5 sha256:2e1c130fa3ec1777a82123374b4c500623959f903c1dd731ee4a83e1f1b38ff2 1.05MB / 3.14MB 1.5s
#5 sha256:2e1c130fa3ec1777a82123374b4c500623959f903c1dd731ee4a83e1f1b38ff2 2.10MB / 3.14MB 2.3s
#5 sha256:84c8c79126f669beec1dcf6f34cd88094471745570c19c29b465dfa7db1fdabd 243B / 243B 2.1s done
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 0B / 27.14MB 0.2s
#5 sha256:2e1c130fa3ec1777a82123374b4c500623959f903c1dd731ee4a83e1f1b38ff2 3.14MB / 3.14MB 3.0s done
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 1.05MB / 11.04MB 3.2s
#5 sha256:824416e234237961c9c5d4f41dfe5b295a3c35a671ee52889bfb08d8e257ec4c 1.05MB / 2.78MB 3.8s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 2.10MB / 11.04MB 3.6s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 3.15MB / 11.04MB 4.4s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 4.19MB / 11.04MB 5.0s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 5.24MB / 11.04MB 5.9s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 2.10MB / 27.14MB 3.8s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 6.29MB / 11.04MB 6.6s
#5 sha256:824416e234237961c9c5d4f41dfe5b295a3c35a671ee52889bfb08d8e257ec4c 2.10MB / 2.78MB 7.1s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 7.34MB / 11.04MB 7.5s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 8.39MB / 11.04MB 8.1s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 4.19MB / 27.14MB 6.5s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 9.44MB / 11.04MB 8.7s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 10.49MB / 11.04MB 9.3s
#5 sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 11.04MB / 11.04MB 9.6s done
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 6.29MB / 27.14MB 8.1s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 8.39MB / 27.14MB 9.3s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 10.49MB / 27.14MB 10.2s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 12.58MB / 27.14MB 11.1s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 14.68MB / 27.14MB 11.9s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 16.78MB / 27.14MB 12.9s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 18.87MB / 27.14MB 14.0s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 20.97MB / 27.14MB 15.0s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 23.07MB / 27.14MB 16.2s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 25.17MB / 27.14MB 17.1s
#5 sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 27.14MB / 27.14MB 17.8s done
#5 extracting sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3
#5 extracting sha256:8b91b88d557765cd8c6802668755a3f6dc4337b6ce15a17e4857139e5fc964f3 0.6s done
#5 extracting sha256:824416e234237961c9c5d4f41dfe5b295a3c35a671ee52889bfb08d8e257ec4c 0.1s done
#5 extracting sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d
#5 extracting sha256:8d53da26040835f622504d7762fad14d226ac414efeb5363f5febebc89ff224d 0.2s done
#5 DONE 20.9s

#5 [1/6] FROM docker.io/library/python:3.9-slim-buster@sha256:320a7a4250aba4249f458872adecf92eea88dc6abd2d76dc5c0f01cac9b5399
0
#5 extracting sha256:84c8c79126f669beec1dcf6f34cd88094471745570c19c29b465dfa7db1fdabd 0.0s done
#5 extracting sha256:2e1c130fa3ec1777a82123374b4c500623959f903c1dd731ee4a83e1f1b38ff2 0.1s done
#5 DONE 21.1s

#6 [2/6] WORKDIR /app
```

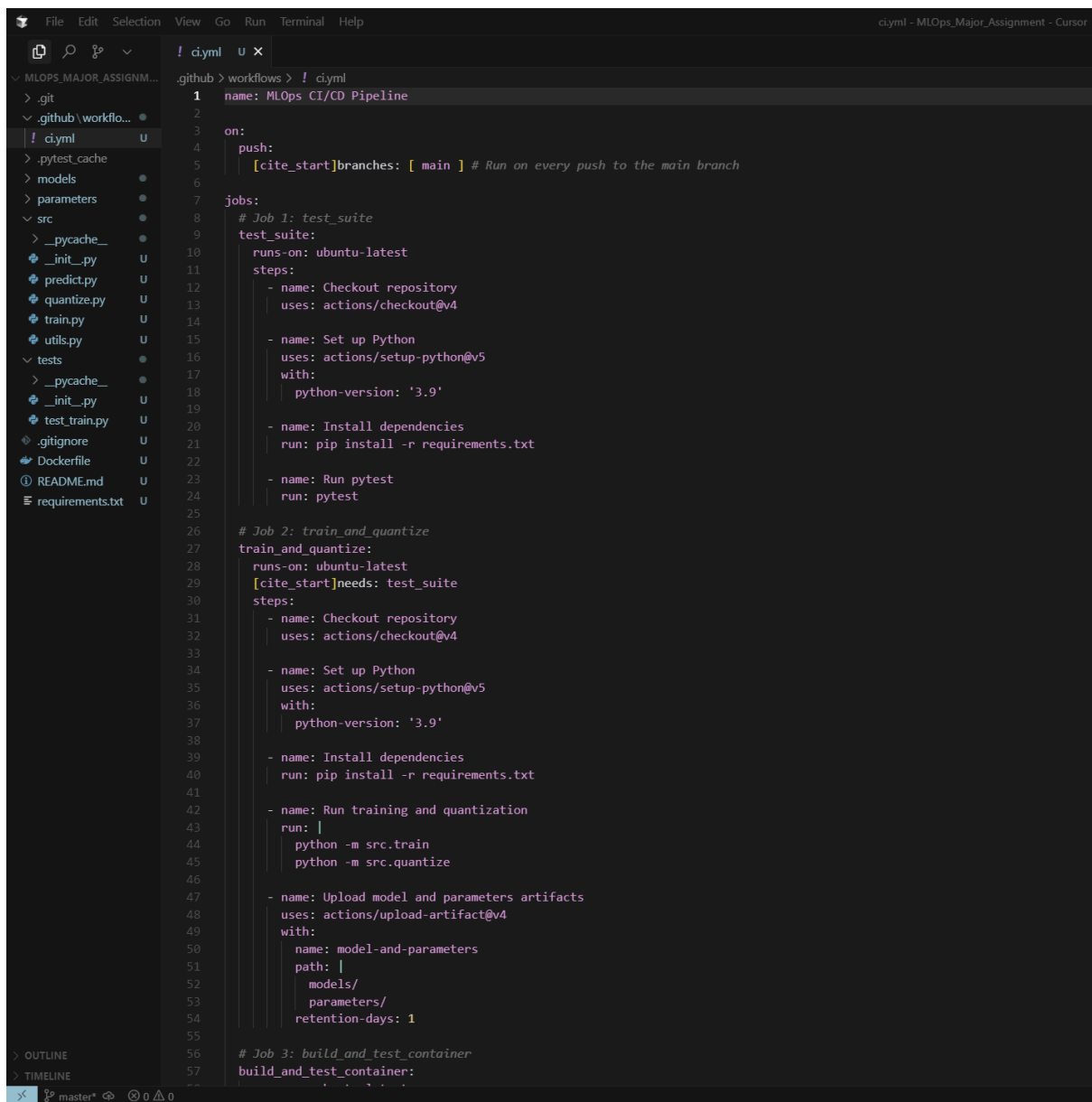
Phase 6 : CI/CD Workflow (.github/workflows/ci.yml)

Step 1: Create the directories and file

GitBash Command: mkdir -p .github/workflows

touch .github/workflows/ci.yml

Step 2: Add the YAML code to ci.yml



```
1 name: MLOps CI/CD Pipeline
2
3 on:
4   push:
5     [cite_start]branches: [ main ] # Run on every push to the main branch
6
7 jobs:
8   # Job 1: test_suite
9   test_suite:
10     runs-on: ubuntu-latest
11     steps:
12       - name: Checkout repository
13         uses: actions/checkout@v4
14
15       - name: Set up Python
16         uses: actions/setup-python@v5
17         with:
18           python-version: '3.9'
19
20       - name: Install dependencies
21         run: pip install -r requirements.txt
22
23       - name: Run pytest
24         run: pytest
25
26   # Job 2: train_and_quantize
27   train_and_quantize:
28     runs-on: ubuntu-latest
29     [cite_start]needs: test_suite
30     steps:
31       - name: Checkout repository
32         uses: actions/checkout@v4
33
34       - name: Set up Python
35         uses: actions/setup-python@v5
36         with:
37           python-version: '3.9'
38
39       - name: Install dependencies
40         run: pip install -r requirements.txt
41
42       - name: Run training and quantization
43         run: |
44           python -m src.train
45           python -m src.quantize
46
47       - name: Upload model and parameters artifacts
48         uses: actions/upload-artifact@v4
49         with:
50           name: model-and-parameters
51           path: |
52             models/
53             parameters/
54           retention-days: 1
55
56   # Job 3: build_and_test_container
57   build_and_test_container:
```

Step 3: Commit and Push to GitHub

GitBash Command: git add .

git commit -m "feat: Add CI/CD workflow, Dockerfile, and predict.py"

git push origin main

```
AKSHAY@Akshay MINGW64 ~/OneDrive/Desktop/IITJ/Semester 3/ML Ops/Major Assignment/MLOps_Major_Assignment (main)
$ git push -u origin main
Enumerating objects: 31, done.
Counting objects: 100% (31/31), done.
Delta compression using up to 16 threads
Compressing objects: 100% (27/27), done.
Writing objects: 100% (31/31), 15.95 KiB | 2.66 MiB/s, done.
Total 31 (delta 1), reused 0 (delta 0), pack-reused 0
remote: Resolving deltas: 100% (1/1), done.
To https://github.com/Akshaykumarky26/MLOps_Major_Assignment.git
* [new branch]      main -> main
branch 'main' set up to track 'origin/main'.
(mlops_env)
```

CONCLUSION

We performed quantization-aware regression using 8-bit integers and preserved the intercept in float precision. After trying multiple quantization strategies (global, per-coefficient, symmetric scaling), we found that global symmetric quantization with clipped small coefficients gave the best balance between model size and accuracy. The final quantized model achieved an R^2 score of 0.4376 (original: 0.5758), with acceptable prediction quality.

Model Comparison Table

Metric	Original Model	Quantized Model
R ² Score	0.5758	0.4376
Mean Squared Error	0.5559	0.7370
File Size	681 Bytes	403 Bytes

feat: Add CI/CD workflow, Dockerfile, and predict.py #4

Re-run all jobs

Summary

Jobs

test_suite

train_and_quantize

build_and_test_container

Run details

Usage

Workflow file

Triggered via push 4 minutes ago

Akshayky26 pushed · 8b820cf main

Status

Success

Total duration

1m 18s

Artifacts

1

ci.yml

on: push

test_suite

22s

train_and_quantize

22s

build_and_test_container

25s