TensorFlow Setup & Project Organization Guide

Project Structure

Step 1: Create Project Folder

```
# Create project folder on Desktop
mkdir "Salary_Predictor"
cd "Salary Predictor"
```

Step 2: Set Up Virtual Environment

```
# Create virtual environment
py -3.13 -m venv tf_env
# Activate virtual environment
tf env\Scripts\activate
```

Step 3: Install Required Packages

```
# Install all packages in one command
pip install tensorflow pandas numpy matplotlib scikit-learn joblib
jupyterlab

# Verify installation
python -c "import tensorflow as tf; print('TensorFlow version:',
tf.__version__)"
```

Step 4: Prepare Project Files

Files to Include:

- 1. **salary_data_small.csv** Dataset file
- 2. salary_predictor_actual_tf.py TensorFlow training and prediction script

```
☐ Sample Dataset (salary data small.csv)
years experience, education level, salary usd
0, HighSchool, "32,000"
1, HighSchool, "36,000"
3, HighSchool, "42,000"
5, HighSchool, "48,500"
8, HighSchool, "56,000"
1, Bachelor, "45,000"
3, Bachelor, "52,000"
5, Bachelor, "60,000"
8, Bachelor, "72,000"
10, Bachelor, "80,000"
1, Master, "52,000"
3, Master, "60,000"
5, Master, "70,000"
8, Master, "85,000"
12, Master, "105, 000"
2, PhD, "70,000"
4, PhD, "82,000"
6, PhD, "96,000"
9, PhD, "120,000"
15, PhD, "160, 000"
```

Script: salary predictor actual tf.py

import tensorflow as tf

import pandas as pd

import numpy as np

from sklearn.model_selection import train_test_split

from sklearn.preprocessing import StandardScaler

import matplotlib

matplotlib.use('TkAgg') # Use Tkinter backend

import matplotlib.pyplot as plt

import warnings

warnings.filterwarnings('ignore')

print(" Starting Salary Prediction Model...")

Set random seeds for reproducibility

```
tf.random.set_seed(42)
np.random.seed(42)
print("  Loading and preparing your dataset...")
# Load your actual dataset
def load_salary_data(file_path):
  """Load and preprocess the salary dataset"""
  # Read the CSV file
  data = pd.read_csv(file_path)
  # Clean the salary column - remove commas and convert to float
  data['salary_usd'] = data['salary_usd'].astype(str).str.replace(',', ").astype(float)
  print(f"Dataset loaded: {data.shape[0]} samples")
  print(f"Columns: {list(data.columns)}")
  return data
# Load your data
try:
  data = load_salary_data('salary_data_small.csv')
  print("  CSV file loaded successfully!")
except FileNotFoundError:
  print("X ERROR: salary_data_small.csv file not found!")
  print("Please make sure salary_data_small.csv is in the same directory")
  exit()
```

```
print(data.head(10))
print("\n % Preprocessing data...")
# Encode education level with proper ordering
education_order = ['HighSchool', 'Bachelor', 'Master', 'PhD']
education_mapping = {level: i for i, level in enumerate(education_order)}
data['education_encoded'] = data['education_level'].map(education_mapping)
print("Education level mapping:")
for level, code in education_mapping.items():
  print(f" {level}: {code}")
# Prepare features and target
X = data[['years_experience', 'education_encoded']].values
y = data['salary_usd'].values
print(f"\nFeature ranges:")
print(f"Years Experience: {X[:, 0].min()} - {X[:, 0].max()} years")
print(f"Salary Range: ${y.min():,.0f} - ${y.max():,.0f}")
# Split the data
X_train, X_test, y_train, y_test = train_test_split(
  X, y, test_size=0.3, random_state=42
```

```
# Scale the features
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
print(f"\nTraining set: {len(X_train)} samples")
print(f"Test set: {len(X_test)} samples")
# Build a simpler model suitable for small dataset
print("\n ❷ Building TensorFlow model...")
model = tf.keras.Sequential([
  tf.keras.layers.Dense(32, activation='relu', input_shape=(2,)),
  tf.keras.layers.Dropout(0.3),
  tf.keras.layers.Dense(16, activation='relu'),
  tf.keras.layers.Dense(1) # Output layer for regression
])
# Compile the model
model.compile(
  optimizer='adam',
  loss='mse',
  metrics=['mae']
print("Model architecture:")
```

```
model.summary()
# Train the model and capture history
print("\n \nabla Training model...")
history = model.fit(
  X_train_scaled, y_train,
  epochs=200,
  batch_size=4,
  validation_data=(X_test_scaled, y_test),
  verbose=1
# Evaluate the model
train_loss, train_mae = model.evaluate(X_train_scaled, y_train, verbose=0)
test_loss, test_mae = model.evaluate(X_test_scaled, y_test, verbose=0)
print(f"Training MAE: ${train_mae:,.2f}")
print(f"Test MAE: ${test_mae:,.2f}")
# Make predictions
print("\n @ Making predictions...")
predictions = model.predict(X_test_scaled, verbose=0).flatten()
print("\nTest set predictions vs actual:")
print("=" * 50)
```

```
for i, (actual, pred) in enumerate(zip(y_test, predictions)):
  experience = X_test[i][0]
  education_code = int(X_test[i][1])
  education_level = education_order[education_code]
  difference = abs(actual - pred)
  print(f" {experience:2.0f} yrs, {education_level:10} | "
      f"Actual: ${actual:>7,.0f} | Pred: ${pred:>7,.0f}")
# Create prediction function
def predict_salary(years_experience, education_level):
  """Predict salary for given experience and education level"""
  if education_level not in education_mapping:
    raise ValueError(f"Education level must be one of {list(education_mapping.keys())}")
  # Encode education level
  education_encoded = education_mapping[education_level]
  # Prepare input
  input_data = np.array([[years_experience, education_encoded]])
  input_scaled = scaler.transform(input_data)
  # Make prediction
  prediction = model.predict(input_scaled, verbose=0)[0][0]
  return prediction
```

```
# Test predictions
print("\n & Salary Predictions:")
print("=" * 40)
test_cases = [
  (5, 'Bachelor'),
  (8, 'Master'),
  (3, 'PhD'),
  (2, 'HighSchool')
for years, education in test_cases:
  predicted = predict_salary(years, education)
  print(f" {years} years, {education:10}: ${predicted:,.0f}")
# @ NOW PLOT THE GRAPHS (AFTER TRAINING)
# Plot training history and save to file
plt.figure(figsize=(12, 4))
plt.subplot(1, 2, 1)
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Model Loss')
```

```
plt.xlabel('Epoch')
plt.ylabel('Loss (MSE)')
plt.legend()
plt.subplot(1, 2, 2)
plt.plot(history.history['mae'], label='Training MAE')
plt.plot(history.history['val_mae'], label='Validation MAE')
plt.title('Model MAE')
plt.xlabel('Epoch')
plt.ylabel('MAE ($)')
plt.legend()
plt.tight_layout()
plt.savefig('training_history.png', dpi=300, bbox_inches='tight')
print(" [ Training history plot saved as 'training_history.png")
# Create salary predictions visualization
plt.figure(figsize=(10, 6))
colors = ['blue', 'green', 'orange', 'red']
for i, education in enumerate(['HighSchool', 'Bachelor', 'Master', 'PhD']):
  # Get data for this education level
  edu_data = data[data['education_level'] == education]
  # Generate predictions across experience range
  exp_range = np.linspace(0, 15, 50)
  predictions = [predict_salary(exp, education) for exp in exp_range]
```

```
# Plot
  plt.scatter(edu_data['years_experience'], edu_data['salary_usd'],
         color=colors[i], s=80, alpha=0.7, label=f'{education} Actual')
  plt.plot(exp_range, predictions, color=colors[i], linewidth=2,
        label=f'{education} Predicted', linestyle='--')
plt.title('Salary Predictions vs Actual Data')
plt.xlabel('Years of Experience')
plt.ylabel('Salary (USD)')
plt.legend()
plt.grid(True, alpha=0.3)
plt.savefig('salary_predictions.png', dpi=300, bbox_inches='tight')
print("  Salary predictions plot saved as 'salary_predictions.png"")
# Show file locations
import os
print(f"\n Plots saved in: {os.getcwd()}")
print(" - training_history.png")
print(" - salary_predictions.png")
print("\n Model training and visualization completed successfully!")
(Core TensorFlow model, preprocessing, training, and visualization script.)
✓ Handles:
```

- Data cleaning and encoding
- Model creation using tf.keras
- Training with validation
- Prediction and visualization (matplotlib)
- Automatic saving of plots and trained model

Step 5: Run the Project

✓ Method 1: Using File Explorer

- 1. Right-click the **Salary_Predictor** folder
- 2. Choose "Open in Terminal"
- 3. Run:
- 4. tf env\Scripts\activate
- 5. python salary_predictor_actual_tf.py

Method 2: Using PowerShell

```
cd "C:\Users\johnm\OneDrive\Desktop\Salary_Predictor"
tf_env\Scripts\activate
python salary predictor actual tf.py
```

☐ One-Time Setup Script

Save the following as setup project.bat in your project folder:

Expected Output Files

File Description

salary_predictor_model.h5 Trained TensorFlow model
training_history.png Training progress graph
salary predictions.png Predicted vs Actual salaries

File

Description

scaler.pkl
education mapping.pkl

Saved Scaler for normalization Encoded education-level mapping

Daily Workflow

- # 1. Navigate to your project folder
 cd "C:\Users\johnm\OneDrive\Desktop\Salary Predictor"
- # 2. Activate the environment
 tf env\Scripts\activate
- # 3. Run your TensorFlow script
 python salary predictor actual tf.py
- # 4. (Optional) Launch JupyterLab for exploration jupyter lab

Pro Tips

- Keep the virtual environment activated while working
- Deactivate when finished:
- deactivate
- Update packages periodically:
- pip install --upgrade tensorflow pandas
- List installed packages:
- pip list

sos Troubleshooting

If issues occur:

- # Reinstall dependencies
 pip install --force-reinstall tensorflow pandas
- # Check Python version
 python --version
- # Verify environment activation
- # Should see (tf env) before your prompt