Here's your Full Pytorch Code for Indian

housing price prediction, to accept user input after

training — so users can predict prices for their own custom home features (like size, location, bedrooms, and bathrooms).

✓ Summary

Step	Description
Imports	Loads required ML and data tools
Simulated Data	Creates fake Indian housing dataset
Preprocessing	Encodes, scales, splits the data
Neural Network	Defines and trains a PyTorch model
Evaluation	Measures model accuracy on test data

Same example usign csv data file :-

Sample CSV Data: housing_data.csv

location, size, bedrooms, bathrooms, price Mumbai, 1200, 2, 2, 8500000 Delhi, 1500, 3, 2, 9500000 Bangalore, 1300, 2, 2, 7800000 Chennai, 1100, 2, 1, 6200000 Kolkata, 1000, 2, 1, 5000000 Mumbai, 2000, 3, 3, 14000000 Delhi, 1800, 3, 2, 11500000 Bangalore, 1600, 3, 2, 9200000 Chennai, 1700, 3, 2, 8700000 Kolkata, 1400, 3, 2, 7300000 Mumbai, 2500, 4, 3, 17000000 Delhi, 2400, 4, 3, 16000000 Bangalore, 2200, 4, 3, 13500000 Chennai, 2100, 4, 3, 12800000 Kolkata, 1900, 4, 3, 10500000 Mumbai, 800, 1, 1, 4800000 Delhi, 900, 1, 1, 5200000 Bangalore, 950, 1, 1, 5100000 Chennai, 1000, 1, 1, 4700000

```
Kolkata,850,1,1,4300000

Mumbai,1450,2,2,9200000

Delhi,1350,2,2,8800000

Bangalore,1550,2,2,8900000

Chennai,1250,2,2,8100000

Kolkata,1150,2,2,7500000
```

How to Use housing data.csv:

- 1. Open any text/code editor or Excel.
- 2. Copy-paste the content above.
- 3. Save the file as: housing_data.csv
- 4. Place it in the **same folder** as your .py file running the PyTorch model.

Full python code pytorch-housing-ai-model.py:-

```
# Step 1: Imports
import torch
import torch.nn as nn
import torch.optim as optim
import pandas as pd
import numpy as np
from sklearn.model selection import train test split
from sklearn.preprocessing import StandardScaler
# Step 2: Load Housing Data from CSV
np.random.seed(42)
torch.manual_seed(42)
# 덛 Load your CSV file here
df = pd.read_csv("housing_data.csv") # Make sure this file exists in your
working directory
# Store the original location list for user input
locations = df['location'].unique().tolist()
# Step 3: Preprocessing
df = pd.get_dummies(df, columns=['location'], drop_first=True)
X = df.drop('price', axis=1).values
y = df['price'].values.reshape(-1, 1)
scaler_X = StandardScaler()
scaler y = StandardScaler()
```

```
X = scaler X.fit transform(X)
y = scaler y.fit transform(y)
# Convert to PyTorch tensors
X = torch.tensor(X, dtype=torch.float32)
y = torch.tensor(y, dtype=torch.float32)
# Step 4: Split into train and test sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)
# Step 5: Define the Neural Network
class HousingModel(nn.Module):
    def __init__(self, input_dim):
        super(HousingModel, self). init ()
        self.model = nn.Sequential(
            nn.Linear(input_dim, 128),
            nn.ReLU(),
            nn.Linear(128, 64),
            nn.ReLU(),
            nn.Linear(64, 1)
    def forward(self, x):
        return self.model(x)
# Step 6: Instantiate the model, loss function, optimizer
model = HousingModel(X.shape[1])
criterion = nn.MSELoss()
optimizer = optim.Adam(model.parameters(), lr=0.01)
# Step 7: Training loop
epochs = 100
for epoch in range(epochs):
   model.train()
    optimizer.zero_grad()
    outputs = model(X_train)
    loss = criterion(outputs, y_train)
    loss.backward()
    optimizer.step()
   if (epoch+1) % 10 == 0:
        print(f"Epoch {epoch+1}/{epochs}, Loss: {loss.item():.4f}")
# Step 8: Evaluate the model
model.eval()
with torch.no grad():
```

```
predictions = model(X_test)
    test loss = criterion(predictions, y test).item()
# Inverse scale predictions and actual values
preds real = scaler y.inverse transform(predictions.numpy())
y_test_real = scaler_y.inverse_transform(y_test.numpy())
# Step 9: Show Results
print("\n[ Sample Predictions:")
for i in range(5):
    print(f"Predicted: ₹{int(preds_real[i][0])} | Actual:
₹{int(y test real[i][0])}")
print(f"\n Final Test Loss (MSE): {test_loss:.4f}")
# Step 10: Take User Input for Prediction
print("\n@ Enter details to predict housing price:")
# --- Collect user input ---
user size = float(input("Enter house size (in sq ft): "))
user_bedrooms = int(input("Enter number of bedrooms: "))
user_bathrooms = int(input("Enter number of bathrooms: "))
print("Choose location from:", ", ".join(locations))
user_location = input("Enter location: ").strip()
if user_location not in locations:
    raise ValueError(" X Invalid location selected.")
# --- Create input feature vector ---
input_data = {
    'size': user_size,
    'bedrooms': user_bedrooms,
    'bathrooms': user_bathrooms
# Add one-hot encoding for location
for col in df.columns:
    if col.startswith("location "):
        location_name = col.split("_")[1]
        input_data[col] = 1 if user_location == location name else 0
# If location was dropped first, ensure it is handled
if f"location_{user_location}" not in input_data and
f"location_{locations[0]}" not in df.columns:
    # First location is implicitly encoded, so handle it manually
    for loc in locations[1:]:
       input data[f"location {loc}"] = 0
```

```
# Convert to DataFrame for processing
input_df = pd.DataFrame([input_data])
input_scaled = scaler_X.transform(input_df.values)

# Convert to tensor
input_tensor = torch.tensor(input_scaled, dtype=torch.float32)

# --- Make prediction ---
model.eval()
with torch.no_grad():
    prediction = model(input_tensor)
    price_pred = scaler_y.inverse_transform(prediction.numpy())[0][0]

print(f"\n\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt{\sqrt
```

Output:-

```
PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

✓ Final Test Loss (MSE): 0.0264

♠ Enter details to predict housing price:
Enter house size (in sq ft): 400
Enter number of bedrooms: 2
Enter number of bathrooms: 2
Choose location from: Mumbai, Delhi, Bangalore, Chennai, Kolkata Enter location: Mumbai

⑤ Predicted House Price: ₹5972760
```

Sure! Here's a **step-by-step explanation** of your code that builds a **housing price prediction model using PyTorch**, using data loaded from a CSV file.

Step-by-Step Explanation

Step 1: Import Required Libraries

```
import torch
import torch.nn as nn
import torch.optim as optim
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
```

- **torch**: Core PyTorch package for tensors and computation.
- **torch.nn**: Used to define the neural network layers.
- **torch.optim**: Optimizers like Adam or SGD for training.
- pandas: To load and handle tabular data from CSV.
- **numpy**: For numeric operations.
- **sklearn.model_selection.train_test_split**: To split data into training and testing sets.
- **sklearn.preprocessing.StandardScaler**: To scale (normalize) features.

✓ Step 2: Load Housing Data from CSV

```
np.random.seed(42)
torch.manual_seed(42)
df = pd.read_csv("housing_data.csv")
locations = df['location'].unique().tolist()
```

- Sets random seeds for reproducibility.
- Loads housing data from a CSV file.
- Extracts a list of all unique locations for later user input validation.

✓ Step 3: Preprocessing the Data

```
df = pd.get_dummies(df, columns=['location'], drop_first=True)
```

- Converts the **categorical location column** into multiple binary columns using one-hot encoding.
- drop_first=True avoids multicollinearity by dropping one location column (treated as base location).

```
X = df.drop('price', axis=1).values
y = df['price'].values.reshape(-1, 1)
```

- x: Features (size, bedrooms, bathrooms, and location columns).
- y: Target variable (price).

```
scaler_X = StandardScaler()
scaler_y = StandardScaler()
X = scaler_X.fit_transform(X)
y = scaler y.fit transform(y)
```

• Standardizes both input features and target prices (mean = 0, std = 1) for better training performance.

```
X = torch.tensor(X, dtype=torch.float32)
y = torch.tensor(y, dtype=torch.float32)
```

• Converts the scaled NumPy arrays to PyTorch tensors.

✓ Step 4: Split Data into Training and Testing Sets

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random state=42)
```

• 80% of data is used for training, 20% for testing.

✓ Step 5: Define the Neural Network

- Custom PyTorch model with:
 - Input layer \rightarrow 128 neurons \rightarrow ReLU
 - \circ 128 → 64 neurons → ReLU
 - \circ 64 \rightarrow 1 output neuron (predicts price)
- nn.Sequential chains layers.

✓ Step 6: Initialize Model, Loss Function, and Optimizer

```
model = HousingModel(X.shape[1])
criterion = nn.MSELoss()
optimizer = optim.Adam(model.parameters(), lr=0.01)
```

- **Model** takes number of input features (depends on how many one-hot encoded columns).
- Loss Function: Mean Squared Error for regression.
- **Optimizer**: Adam optimizer with a learning rate of 0.01.

✓ Step 7: Train the Model

```
for epoch in range(epochs):
    ...
```

- For 100 epochs:
 - o Set model to training mode.
 - o Clear gradients.
 - o Forward pass (model (X train)).
 - o Compute loss.
 - o Backpropagation (loss.backward()).
 - o Update model weights (optimizer.step()).
 - o Print loss every 10 epochs.

✓ Step 8: Evaluate the Model

```
model.eval()
with torch.no_grad():
    predictions = model(X_test)
```

- Turns off gradient tracking for evaluation.
- Predicts prices on the test set.
- Calculates MSE loss on actual vs predicted values.

```
preds_real = scaler_y.inverse_transform(predictions.numpy())
y_test_real = scaler_y.inverse_transform(y_test.numpy())
```

• Converts scaled predictions and targets back to real INR prices.

✓ Step 9: Display Sample Predictions

```
for i in range(5):
    print(f"Predicted: ₹{int(preds_real[i][0])} | Actual:
₹{int(y test real[i][0])}")
```

• Prints 5 sample predictions compared to actual values.

✓ Step 10: Take User Input and Predict Price

```
user_size = float(input(...))
user_bedrooms = int(input(...))
user_bathrooms = int(input(...))
user_location = input(...).strip()
```

• Collects input from the user for prediction.

```
for col in df.columns:
    if col.startswith("location_"):
        ...
```

• One-hot encodes the user-selected location to match model's input format.

```
input_df = pd.DataFrame([input_data])
input_scaled = scaler_X.transform(input_df.values)
input tensor = torch.tensor(input scaled, dtype=torch.float32)
```

• Formats and scales the input.

```
with torch.no_grad():
    prediction = model(input_tensor)
    price pred = scaler y.inverse transform(prediction.numpy())[0][0]
```

• Predicts the price and converts it back from scaled form.

```
print(f"\n
    Predicted House Price: ₹{int(price pred)}")
```

• Prints the final predicted house price.

✓ Summary

Step What it Does

- 1 Imports libraries
- 2 Loads housing data from CSV
- 3 Preprocesses data (scaling, one-hot encoding)
- 4 Splits data into train/test
- 5 Defines neural network model
- 6 Initializes model, loss, optimizer
- 7 Trains model using MSE loss
- 8 Evaluates on test set
- 9 Shows sample predictions
- 10 Takes user input and predicts housing price