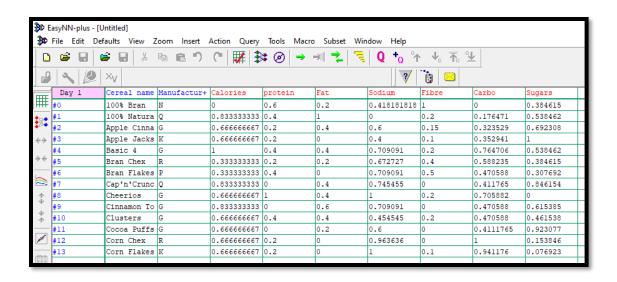
import pandas as pd # Step 3: Split data into training and test sets (80% import numpy as np from sklearn.model_selection import train_test_split training, 20% testing) from sklearn.preprocessing import LabelEncoder X train, X test, y train, y test = train test split(X, y, from tensorflow.keras.models import Sequential test_size=0.2, random_state=42) from tensorflow.keras.layers import Dense from tensorflow.keras.models import load model # Step 4: Create the ANN model model = Sequential() # Step 1: Load the dataset model.add(Dense(units=64, activation='relu', # Assuming you have your dataset saved as a CSV input_dim=X_train.shape[1])) model.add(Dense(units=32, activation='relu')) file. Adjust the path accordingly. # Example DataFrame based on your earlier dataset model.add(Dense(units=1)) # For regression output data = { 'Cereal name': ['100% Bran', '100% Natural Bran', # Step 5: Compile the model 'Apple Cinnamon Cheerios', 'Apple Jacks', model.compile(optimizer='adam', 'Basic 4', 'Bran Chex', 'Bran Flakes', loss='mean_squared_error', metrics=['mae']) 'Cap\'n\'Crunch', 'Cheerios', 'Cinnamon Toast Crunch', 'Clusters', 'Cocoa Puffs', 'Corn Chex', 'Corn Flakes'], # Step 6: Train the model 'Manufact': ['N', 'Q', 'G', 'K', 'G', 'R', 'P', 'Q', 'G', 'G', history = model.fit(X_train, y_train, epochs=100, 'G', 'G', 'R', 'K'], batch_size=10, validation_split=0.2) 'Calories': [0, 0.833333, 0.666667, 0.666667, 1, 0.333333, 0.333333, 0.833333, 0.666667, 0.833333, # Step 7: Evaluate the model 0.666667, 0.666667, 0.666667, 0.666667], loss, mae = model.evaluate(X test, y test) 'Protein': [0.6, 0.4, 0.2, 0.2, 0.4, 0.2, 0.4, 0, 1, 0, print(f"Test Mean Absolute Error: {mae}") 0.4, 0, 0.2, 0.2], 'Fat': [0.2, 1, 0.4, 0, 0.4, 0.2, 0, 0.4, 0.2, 0.6, 0.4, # Step 8: Save the trained model 0.2, 0, 0],model.save('cereal_sugar_prediction_model.h5') 'Sodium': [0.418182, 0.054545, 0.327273, 0.2, print("Model saved as 0.709091, 0.672727, 0.709091, 0.745455, 1, 'cereal sugar prediction model.h5'") 0.709091, 0.454545, 0.327273, 0.963636, 1], 'Fibre': [1, 0.2, 0.15, 0.1, 0.2, 0.4, 0.5, 0, 0.2, 0, 0.2, # Step 9: Load the model (optional) 0, 0, 0], loaded_model = 'Carbo': [0, 0.176471, 0.323529, 0.352941, load model('cereal sugar prediction model.h5') 0.764706, 0.588235, 0.470588, 0.411765, 0.705882, 0.470588, 0.470588, 0.411765, 0.941176, 0.941176],# Step 10: Query the network (predicting sugars for a 'Sugars': [0.384615, 0.538462, 0.692308, new cereal) 0.923077, 0.538462, 0.384615, 0.307692, 0.846154, # Create a sample input (normalized values) 0.076923, 0.615385, 0.538462, 0.692308, 0.153846, sample input = np.array([[0.666667, 1, 0.2, 0.7, 0.4, 0.076923] } 0.7]]) # Replace with actual sample values df = pd.DataFrame(data) # Predict the sugar content using the trained model # Step 2: Preprocess the data predicted sugars = # Convert categorical 'Manufact' column to loaded_model.predict(sample_input) numerical using Label Encoding print(f"Predicted Sugars for sample input: label_encoder = LabelEncoder() {predicted_sugars[0][0]}") df['Manufact'] = label encoder.fit transform(df['Manufact']) # Separate features and target (we'll predict the 'Sugars' column) X = df.drop(columns=['Cereal_name', 'Sugars'])

y = df['Sugars']



1	Cereal nar	Manufacti	Calories	protein	Fat	Sodium	Fibre	Carbo	Sugars
2	100% Bran	N	70	4	1	130	10	5	6
3	100% Natu	Q	120	3	5	15	2	8	8
4	Apple Cin	G	110	2	2	180	1.5	10.5	10
5	Apple Jack	K	110	2	0	125	1	11	14
6	Basic 4	G	130	3	2	210	2	18	8
7	Bran Chex	R	90	2	1	200	4	15	6
8	Bran Flake	P	90	3	0	210	5	13	5
9	Cap'n'Crui	Q	120	1	2	220	0	12	12
10	Cheerios	G	110	6	2	290	2	17	1
11	Cinnamon	G	120	1	3	210	0	13	9
12	Clusters	G	110	3	2	140	2	13	7
13	Cocoa Puf	G	110	1	1	180	0	12	13
14	Corn Chex	R	110	2	0	280	0	22	3
15	Corn Flake	K	100	2	0	290	1	21	2
16									