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course title : CSC 180-01 Intelligent Systems

assignment id : Project\_1

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## Problem Statement

### **Predicting Business Ratings Based on Customer Reviews and Other Business Attributes.**

Given a dataset of businesses with attributes such as customer reviews, location (latitude and longitude), and review count, can we develop a predictive model that accurately estimates the star rating of a business? The model should learn from the provided features and be able to predict star rating of that business.

## Methodology

### **1. Data Gathering and Preprocessing**

- Converted Json data to pandas data frame extracting 1000000 rows. Build two data frames called **review** and **business** Removed businesses with fewer than 20 reviews from the review table.
- Grouped reviews by `business\_id`.
- Merged this Data Frame with the business table using `business\_id`, including relevant features like latitude, longitude, review count, stars, and name.

### **2. Feature Extraction and Transformation**

- Applied TF-IDF vectorization to extract features from review texts.
- Converted the Data Frame into a NumPy array representation.
- Converted additional features (latitude, longitude, review count) to a NumPy array and concatenated it with the TF-IDF features.

### **3. Data Splitting**

- Defined features (X) and target variable (y, stars) and split the dataset into training and testing sets.

#### 4. Model Building

- Developed various models using different activation functions (Relu, Sigmoid, Tanh) and optimizers (Adam, SGD).

#### 5. Evaluation and Visualization

- Plotted graphs to visualize model performance.
- Created a table displaying business names, actual ratings, and predicted ratings.

## Experimental Results and Analysis

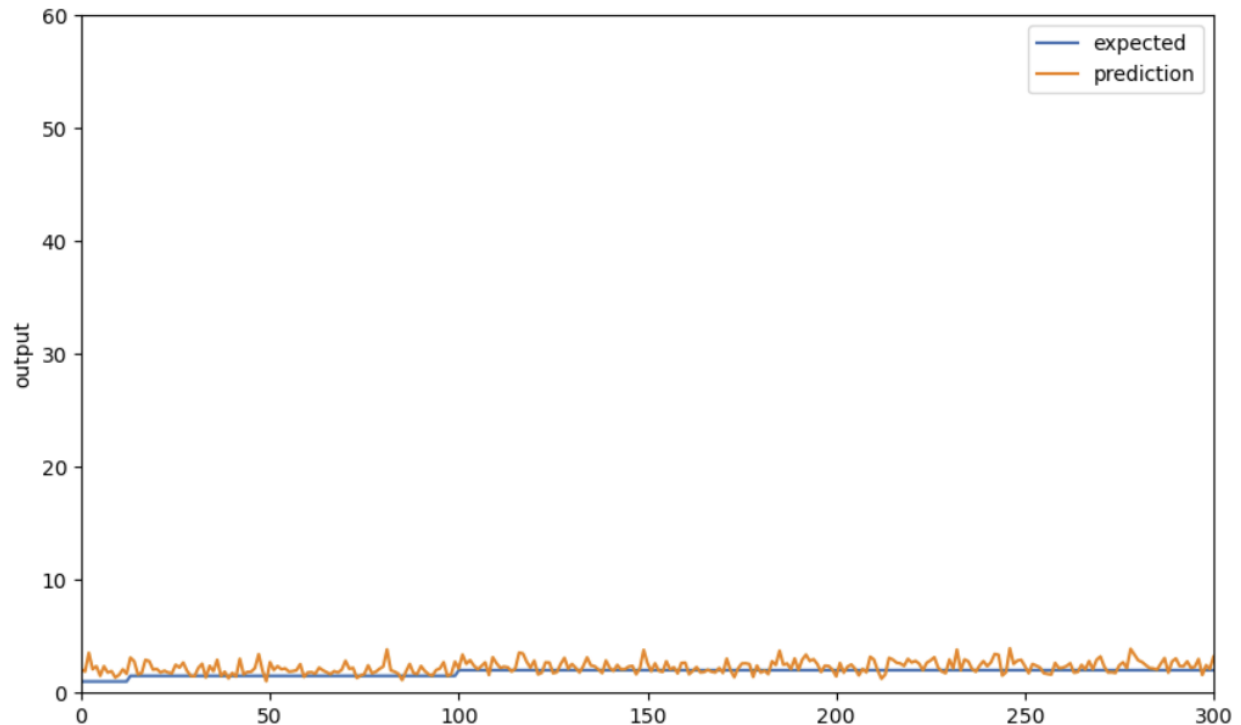
### Model 1

**Model Configuration:** The model was built using a sequential architecture with four hidden layers, utilizing the Relu activation function and the Adam optimizer.

**Performance:** The model achieved an RMSE of approximately 0.4543, indicating that, on average, predicted ratings deviate by about 0.4543 stars from actual ratings.

**Evaluation:** This RMSE value suggests reasonable predictive accuracy, but there is potential for improvement.

	Business Name	True Rating	Predicted Rating
0	Chris's Sandwich Shop	4.5	3.788730
1	Philadelphia	4.0	4.344064
2	Family Vision Center	4.5	4.093851
3	Washoe Metal Fabricating	4.5	4.235305
4	Stewart's De Rooting & Plumbing	4.0	3.751950



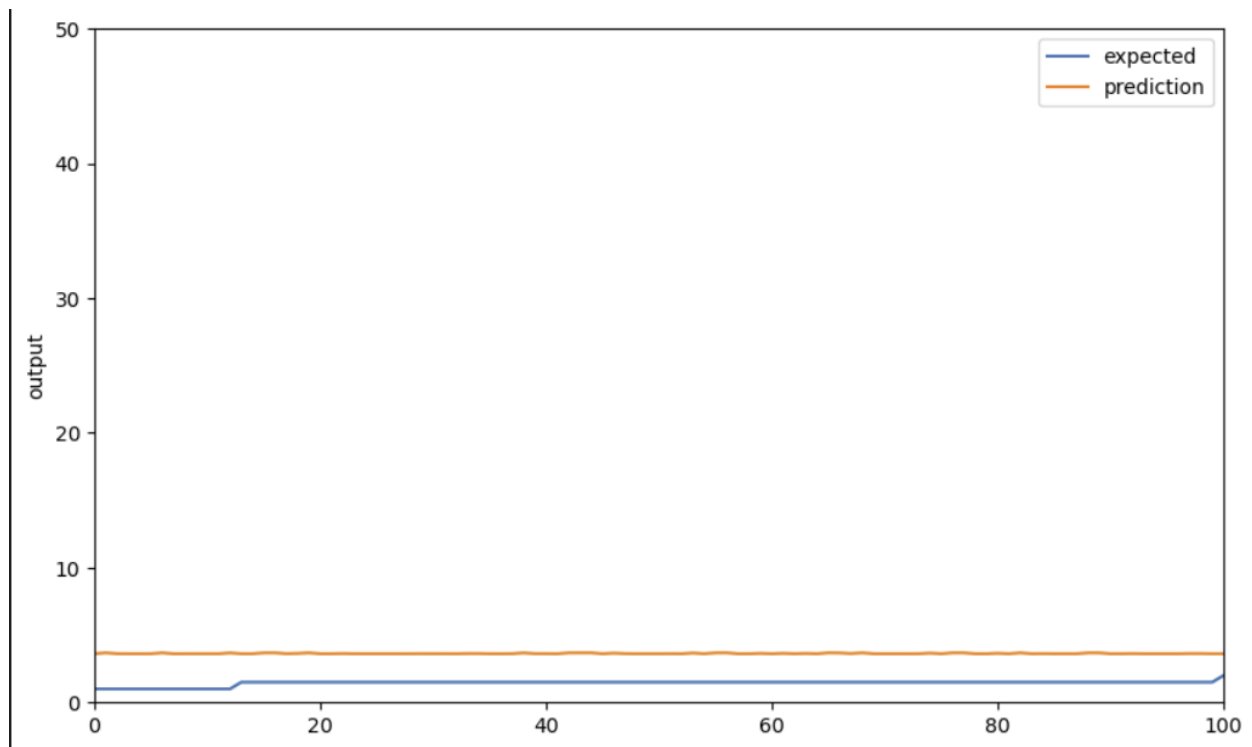
## Model 2

**Model Configuration:** The model was constructed with four hidden layers using the sigmoid activation function and the SGD optimizer.

**Performance:** The model achieved an RMSE of approximately 0.8218, indicating a larger average deviation from actual ratings compared to the previous model.

**Evaluation:** This RMSE suggests lower predictive accuracy, highlighting that the sigmoid activation may not be optimal for this problem.

	Business Name	True Rating	Predicted Rating
0	Chris's Sandwich Shop	4.5	3.635293
1	Philadelphia	4.0	3.679057
2	Family Vision Center	4.5	3.679057
3	Washoe Metal Fabricating	4.5	3.664264
4	Stewart's De Rooting & Plumbing	4.0	3.679057



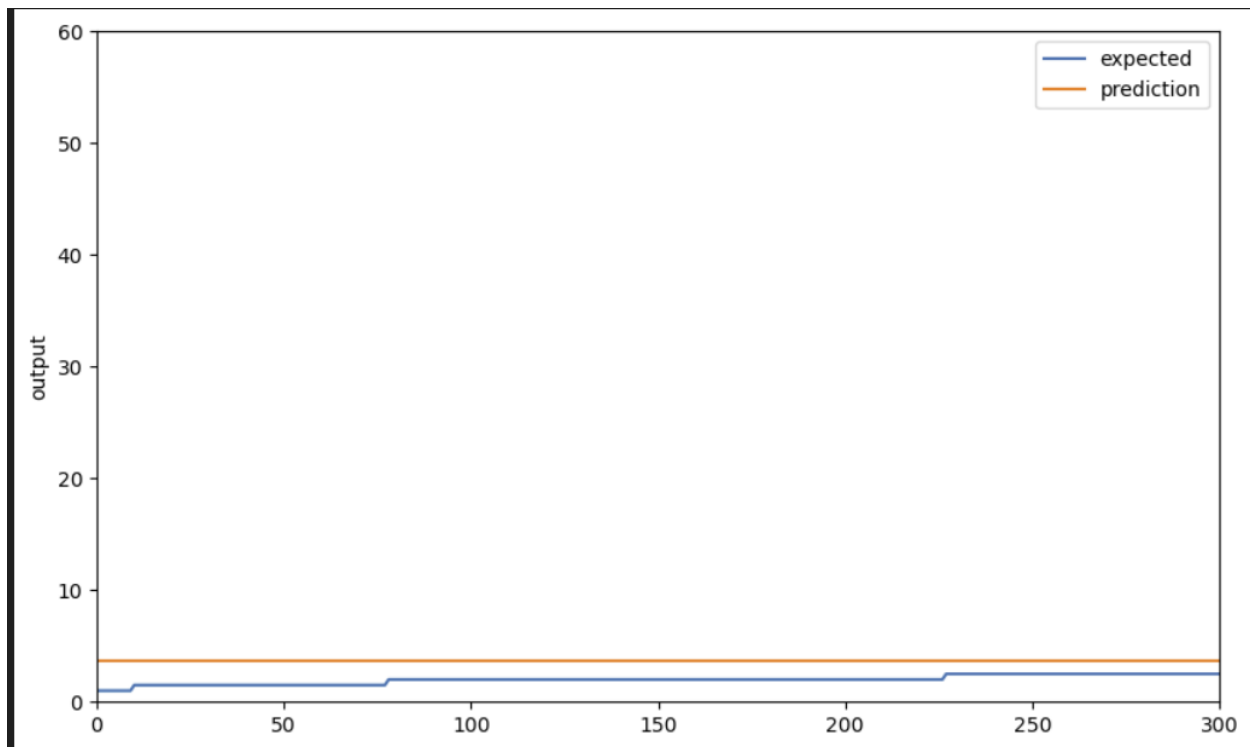
### Model 3

**Model Configuration:** The model consisted of three hidden layers, utilizing the tanh activation function and the Adam optimizer.

**Performance:** The model achieved an RMSE of approximately 0.8323, indicating a further increase in average deviation from actual ratings compared to previous models.

**Evaluation:** This RMSE suggests that the tanh activation may not be the best choice for this dataset, leading to less accurate predictions.

	Business Name	True Rating	Predicted Rating
0	Chris's Sandwich Shop	4.5	3.608639
1	Philadelphia	4.0	3.608639
2	Family Vision Center	4.5	3.608639
3	Washoe Metal Fabricating	4.5	3.608639
4	Stewart's De Rooting & Plumbing	4.0	3.608639



## Task Division and Project Reflection

### Task Division:

- Since the project was completed individually, I was responsible for all aspects, including data collection and preprocessing, feature engineering, model development, and evaluation.

### Challenges Encountered:

- **Chart Creation:** One significant challenge was creating a proper chart for data visualization at the end.

### Learning Outcomes:

- **Concatenating Arrays and Data Frames:** Gained practical experience in combining data structures for model input.
- **Experimenting with Activations and Optimizers:** Learned the effects of different activation functions and optimizers on model performance.
- **Creating Charts:** Developed skills in visualizing data effectively to communicate findings.
- **Understanding Neuron Impact:** Gained insights into how the number of neurons in a model architecture influences predictive capability.