## **Homework 2**

1. Use the dataset from UCI Machine Learning Repository: "Individual household electric power consumption" for performing time series analysis.

https://archive.ics.uci.edu/dataset/235/individual+household+electric+power+consumption

- 2. Perform Exploratory Data Analysis (EDA) of the dataset:
  - Visualize time series trends
  - Check for seasonality and cyclical patterns
  - Analyze distribution of power consumption
  - Identify and handle missing values or outliers
- 3. Implement a linear regression model to predict power consumption for the last three time periods:
  - Split the data into training and testing sets
  - Prepare features (consider lag variables, time-based features)
  - Train the model and make predictions
- 4. Evaluate the linear regression model using appropriate metrics:
  - Mean Absolute Error (MAE)
  - Mean Squared Error (MSE)
  - Root Mean Squared Error (RMSE)
  - R-squared (R2) value
- 5. Implement a Recurrent Neural Network (RNN) for power consumption prediction:
  - Preprocess data for RNN input
  - Design and train the RNN model
  - Make predictions and visualize results
  - Compare performance metrics with linear regression
- 6. Implement Long Short-Term Memory (LSTM) for power consumption prediction:
- Dr. Sharon Yalov-Handzel

## DataScience2 lab

- Preprocess data for LSTM input
- Design and train the LSTM model
- Make predictions and visualize results
- Compare performance metrics with previous models
- 7. Implement an LSTM model with an Attention layer for power consumption prediction:
  - Design and train the LSTM model with Attention
  - Make predictions and visualize results
  - Compare performance metrics with previous models
  - Analyze the Attention weights to interpret model focus
- 8. Data augmentation experiment:
  - Modify up to 10% of the dataset to potentially improve prediction results
  - Retrain and evaluate all three models (RNN, LSTM, LSTM with Attention)
  - Compare the impact of data changes on each model's performance
- 9. Data reduction experiment:
  - Remove up to 10% of the data randomly
  - Retrain and evaluate all three models
  - Compare how data reduction affects each model's performance
- 10. Data resolution experiment:
- Reduce the time resolution of the data by 50% (e.g., from minute-level to 2-minute intervals)
  - Retrain and evaluate all three models
  - Analyze how changes in data resolution impact each model's performance
- 11. Conclusion and insights:
  - Summarize findings from all experiments
  - Discuss which model performed best under different conditions
  - Provide insights on the dataset's characteristics and their impact on model performance
- Dr. Sharon Yalov-Handzel