

Data Generation and Preparation

LSTM Model Build

LSTM Model Training

LSTM Prediction Validation

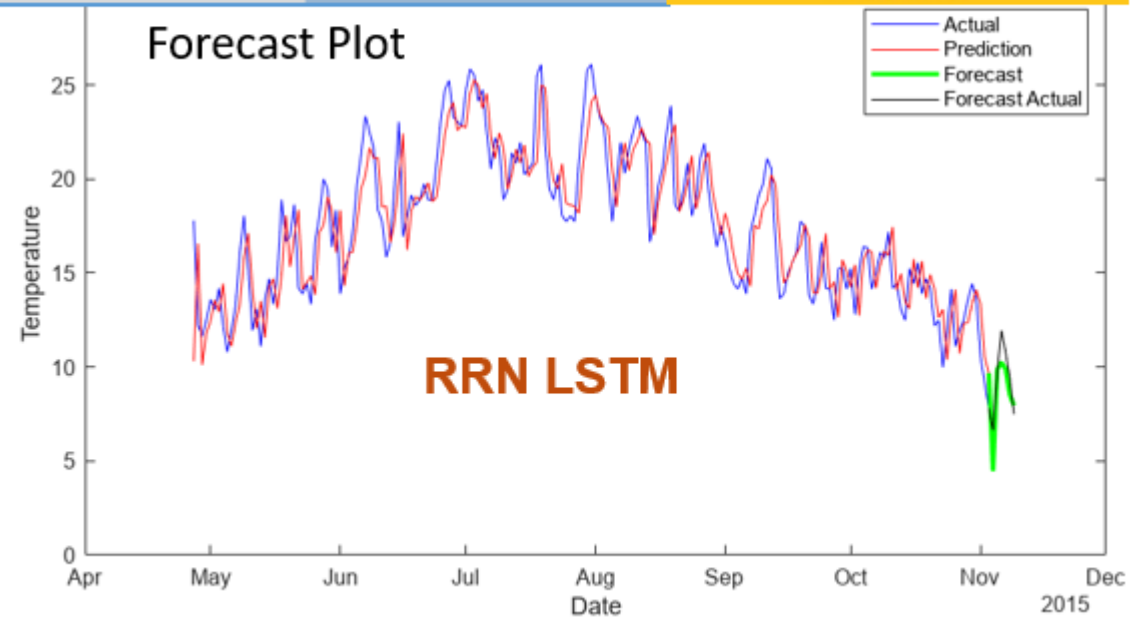
LSTM Forecast Validation

Unsupervised Learning

Unsupervised learning explores unlabeled data to find patterns and structures without explicit guidance or predefined labels.



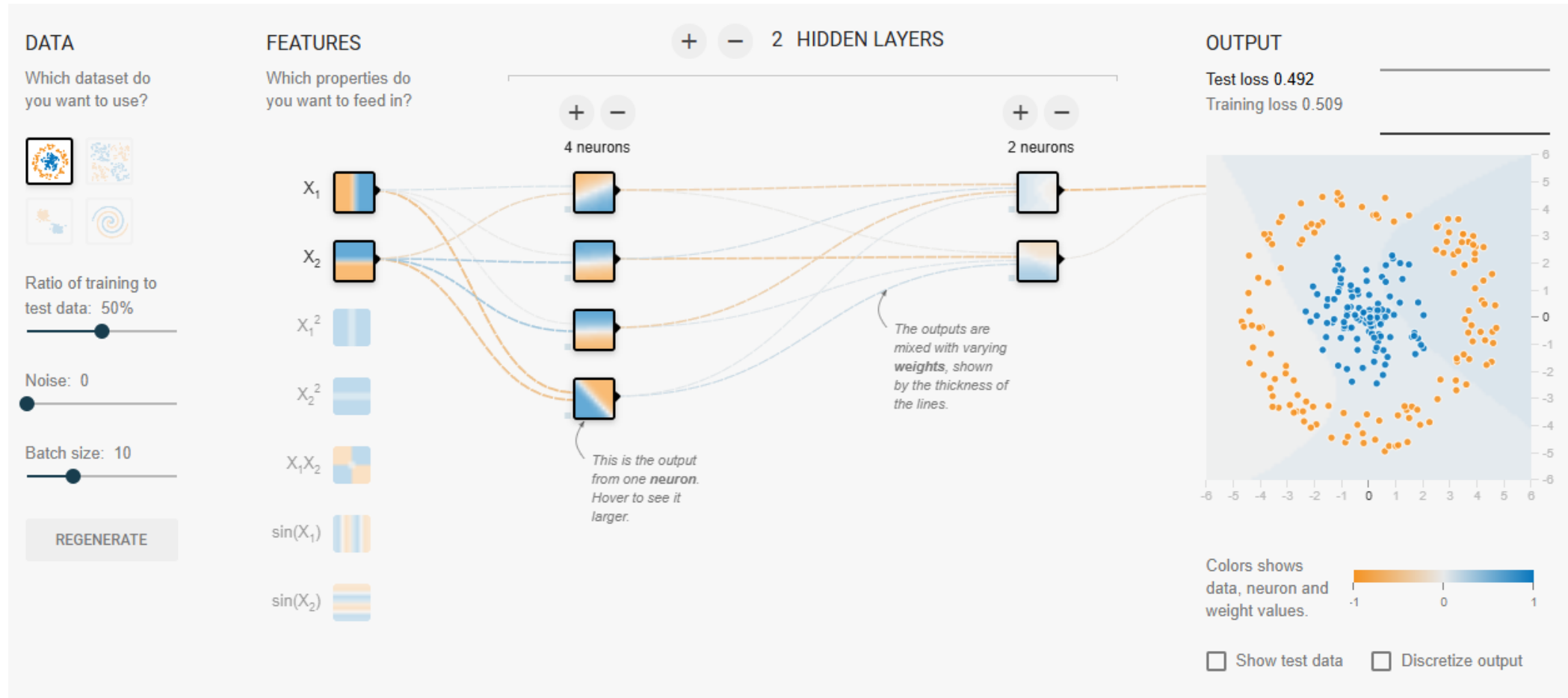
**Control Tech
LAB**



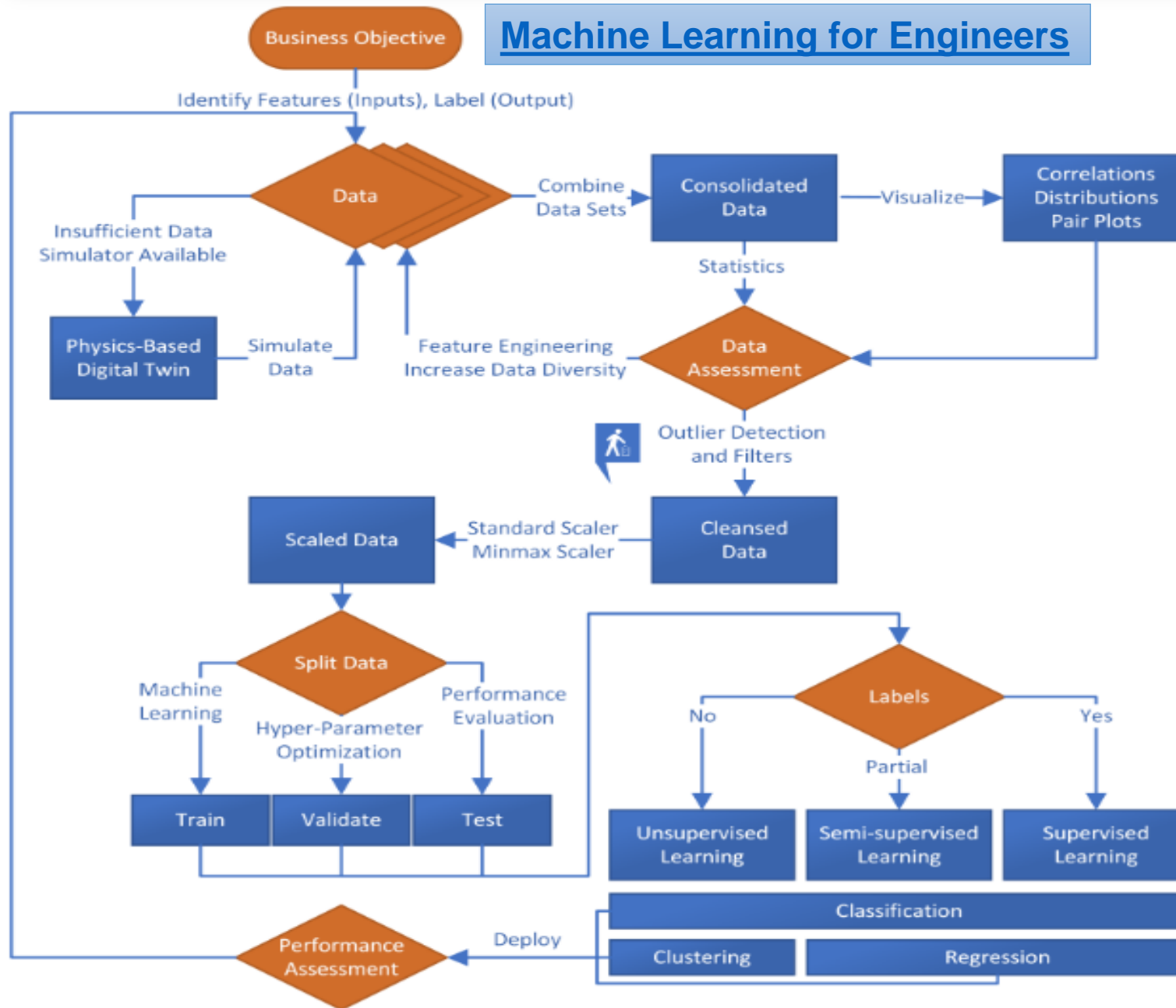
Machine learning

Working with:

1. Introduction to Machine Learning Unsupervised Learning with a Presentation
2. Live Script Python and MATLAB RRN LSTM Model for Prediction and Forecasting Temperature



Machine Learning for Engineers



An Optimized Deep Learning Approach for Forecasting Temperature

CONTEXT

1) Time Series

2) Machine Learning

3) Neural Networks

4) How to Build up ML Models

5) Results and Validation

Machine Learning

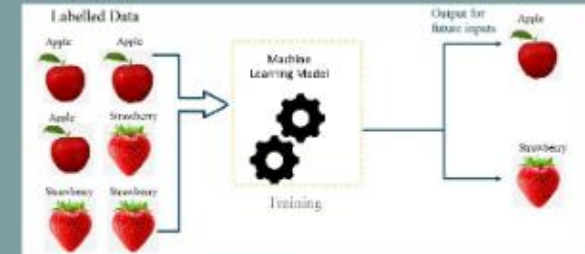
Machine learning is a field of study that focuses on developing algorithms and models that enable computers to learn and make predictions or decisions without being explicitly programmed.

Deep Learning

Deep learning is a subset of machine learning that utilizes neural networks with multiple layers to learn and extract complex patterns from data, enabling accurate predictions and classifications in various domains such as image recognition, speech processing, and natural language understanding.

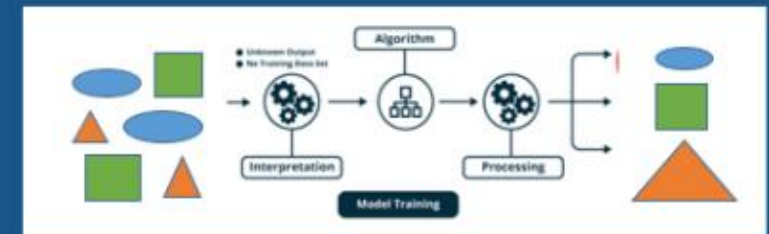
Supervised Learning

Supervised learning uses labeled data to train a model that can make predictions or classifications on new data by learning from the provided labels.



Unsupervised Learning

Unsupervised learning explores unlabeled data to find patterns and structures without explicit guidance or predefined labels.



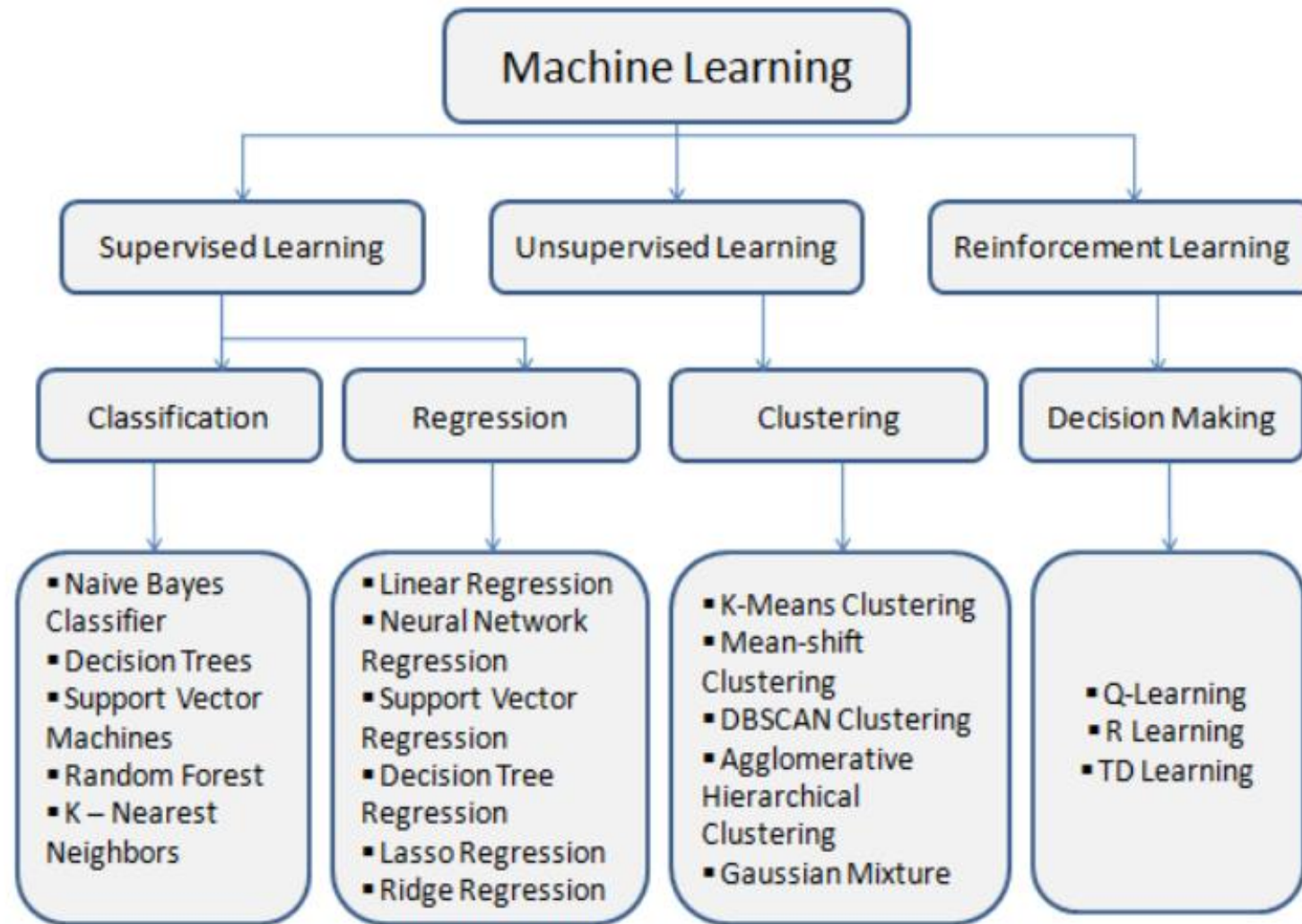
Selecting the Right Algorithm

Selecting the right machine learning algorithm involves balancing trade-offs between flexibility, interpretability, speed, accuracy, and complexity, with no single best method available. Trial and error is essential, as even experienced data scientists need to test different algorithms to find the most effective one.

Model Statistics

Model 2: Tree	
Status: Trained	
Training Results	
RMSE (Validation)	3.2821
R-Squared (Validation)	0.82
MSE (Validation)	10.772
MAE (Validation)	2.3731
Prediction speed	~5000 obs/sec
Training time	3.5947 sec

Statistic	Description	Tip	
RMSE	Root mean squared error. The RMSE is always positive and its units match the units of your response.	Look for smaller values of the RMSE.	$RMSE = \sqrt{\sum_{i=1}^n \frac{(\hat{y}_i - y_i)^2}{n}}$
R-Squared	Coefficient of determination. R-squared is always smaller than 1 and usually larger than 0. It compares the trained model with the model where the response is constant and equals the mean of the training response. If your model is worse than this constant model, then R-Squared is negative.	Look for an R-Squared close to 1.	$R^2 = 1 - \frac{SS_{RES}}{SS_{TOT}} = 1 - \frac{\sum_i (y_i - \hat{y}_i)^2}{\sum_i (y_i - \bar{y})^2}$
MSE	Mean squared error. The MSE is the square of the RMSE.	Look for smaller values of the MSE.	$MSE = \frac{1}{n} \sum_{i=1}^n (y_i - \tilde{y}_i)^2$
MAE	Mean absolute error. The MAE is always positive and similar to the RMSE, but less sensitive to outliers.	Look for smaller values of the MAE.	$MAE = \frac{1}{n} \sum_{i=1}^n Y_i - \hat{Y}_i $



<https://www.analyticsvidhya.com/blog/2021/03/everything-you-need-to-know-about-machine-learning/>

Stationarity: We do not have independence but consistency.

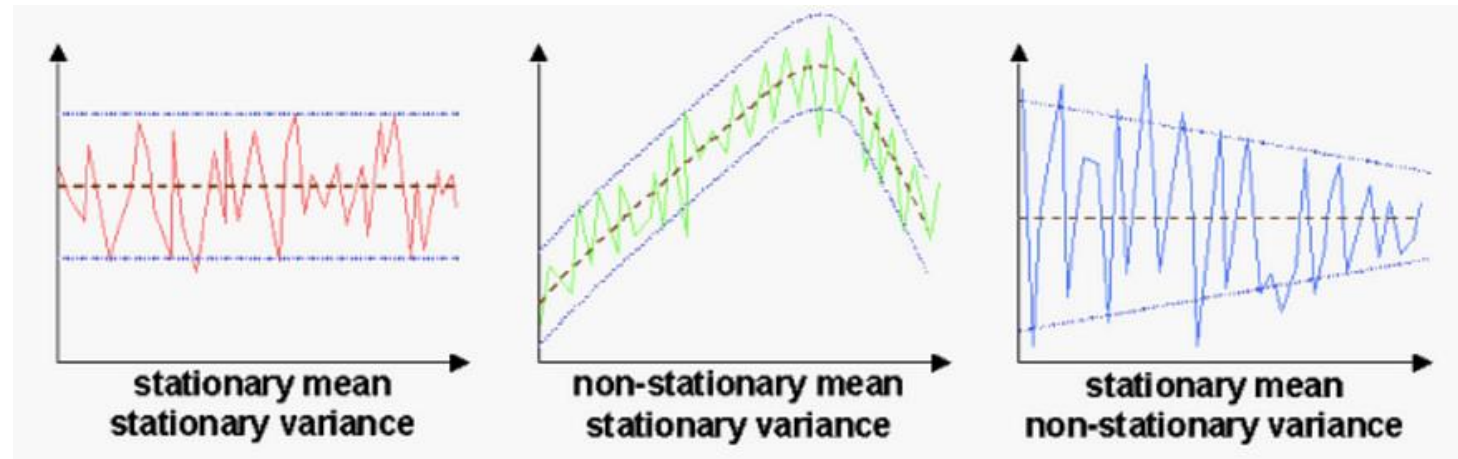
Stationarity is an important concept in time series analysis:

1. Stationarity means that the statistical properties of a time series do not change over time.
2. Stationarity is important because many useful analytical tools, statistical tests and models rely on it.

The ability to determine if a time series is stationary is important. This usually means being able to ascertain, with high probability, that a series is generated by a stationary process.

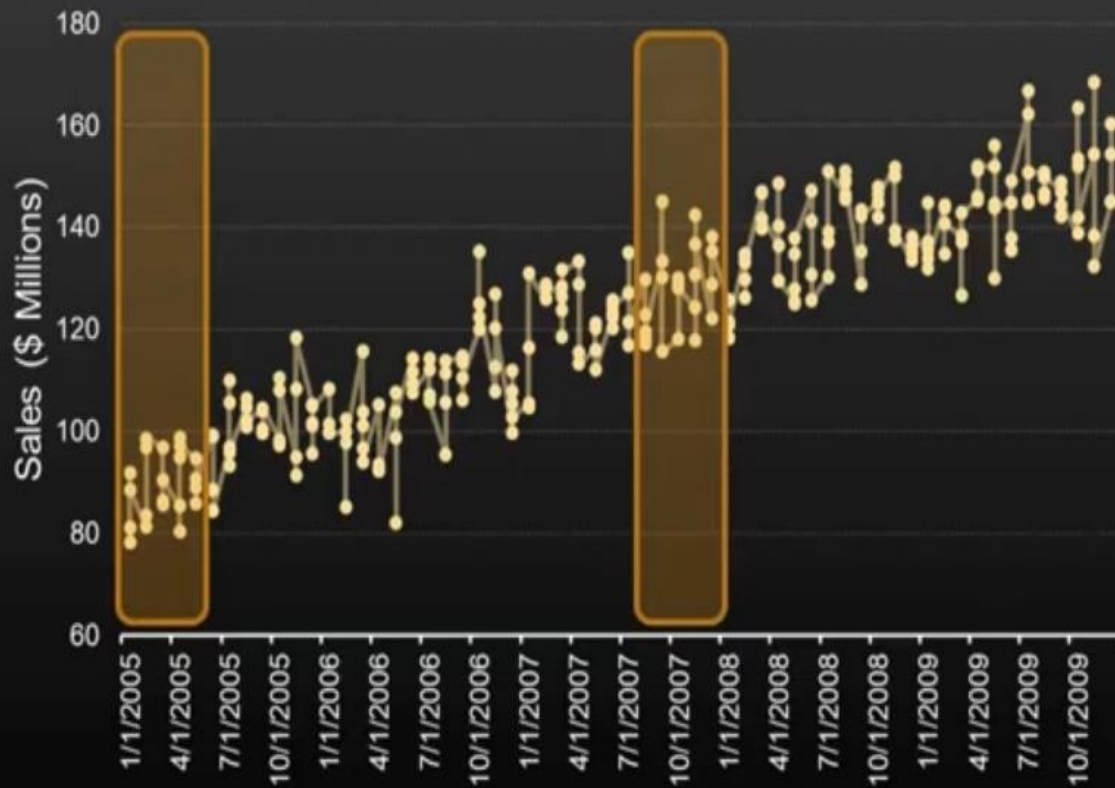
Test stationarity!

Mean, variance, autocorrelation depends **only** on difference in time, **not** location in time.

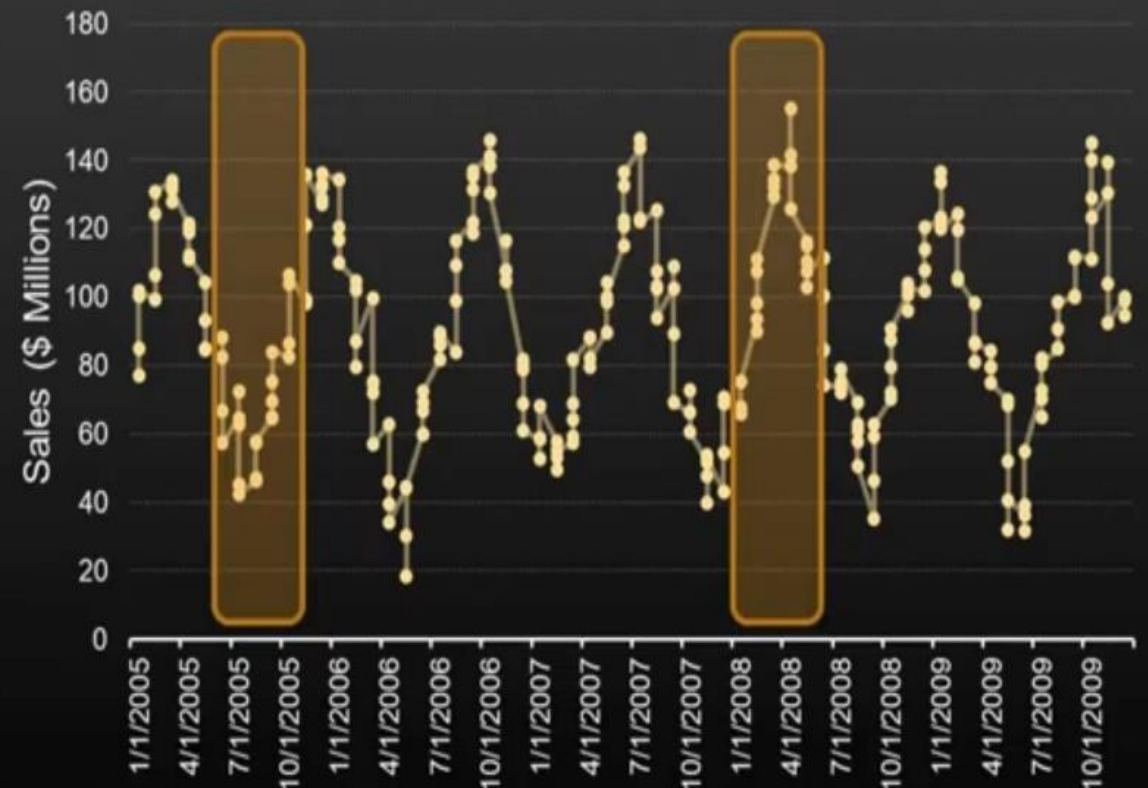


Not stationarity !
They do not have the same mean

Trending



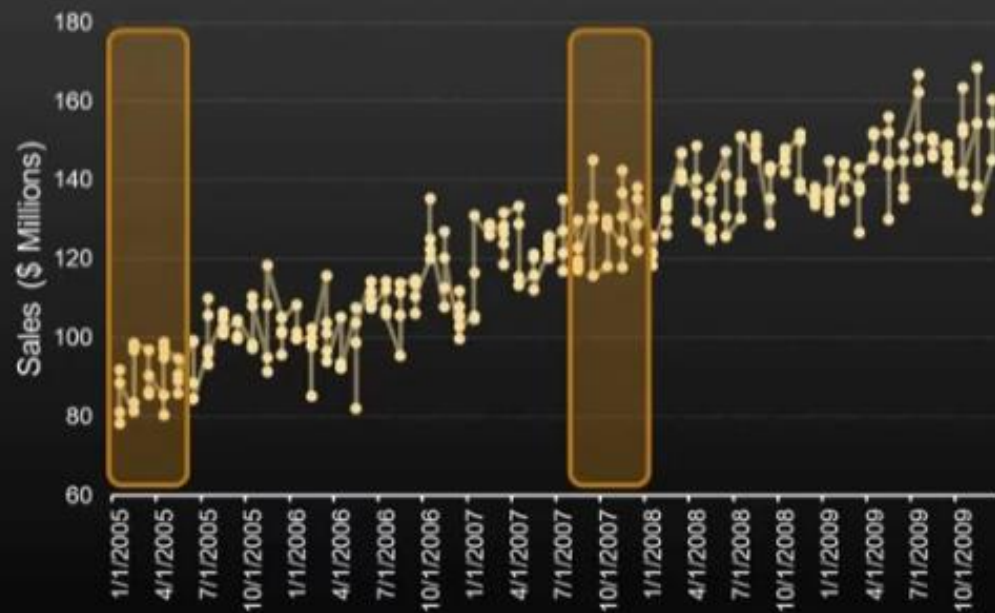
Seasonality



Trending

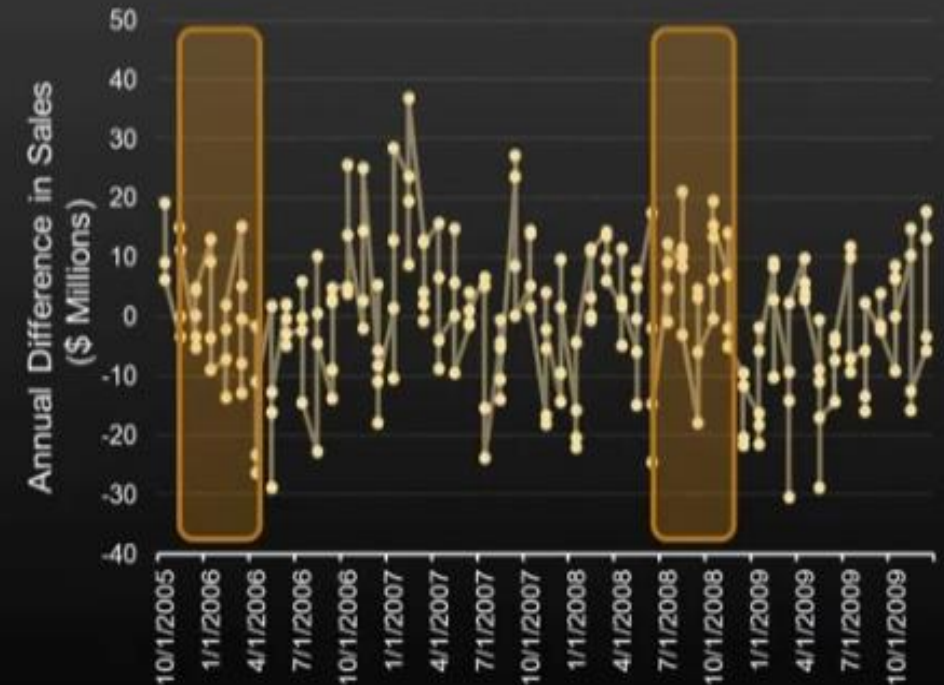
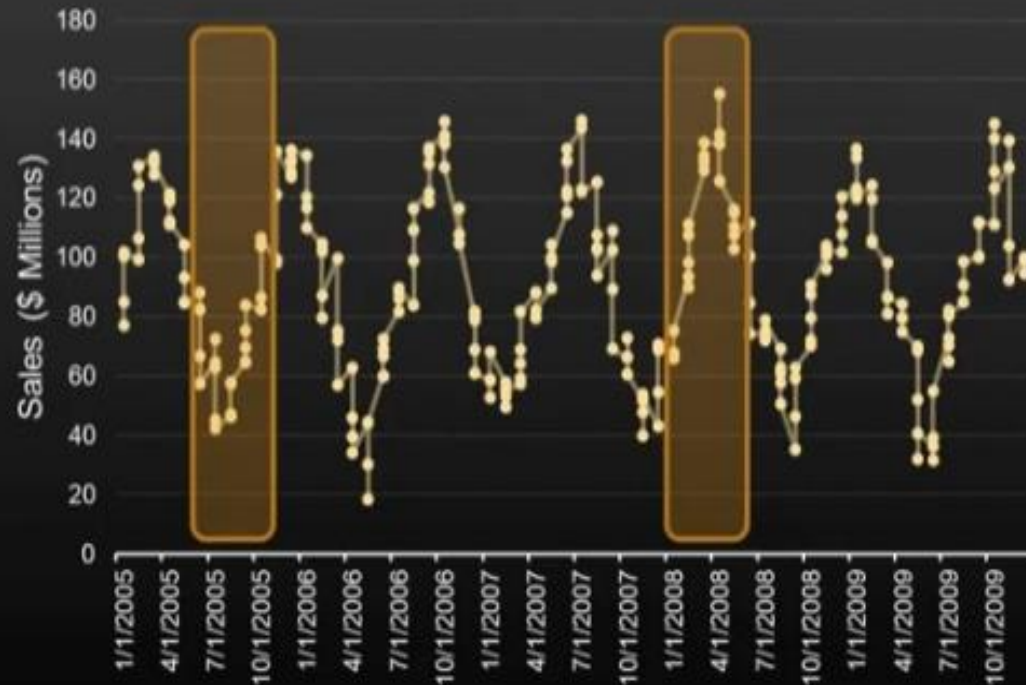
Stationarity!

- **Trend** – look at difference between current point and previous one: $Y_t - Y_{t-1}$



Stationarity!

- **Season** – look at difference between current point and the same point in the previous season: $Y_t - Y_{t-s}$



Correlation

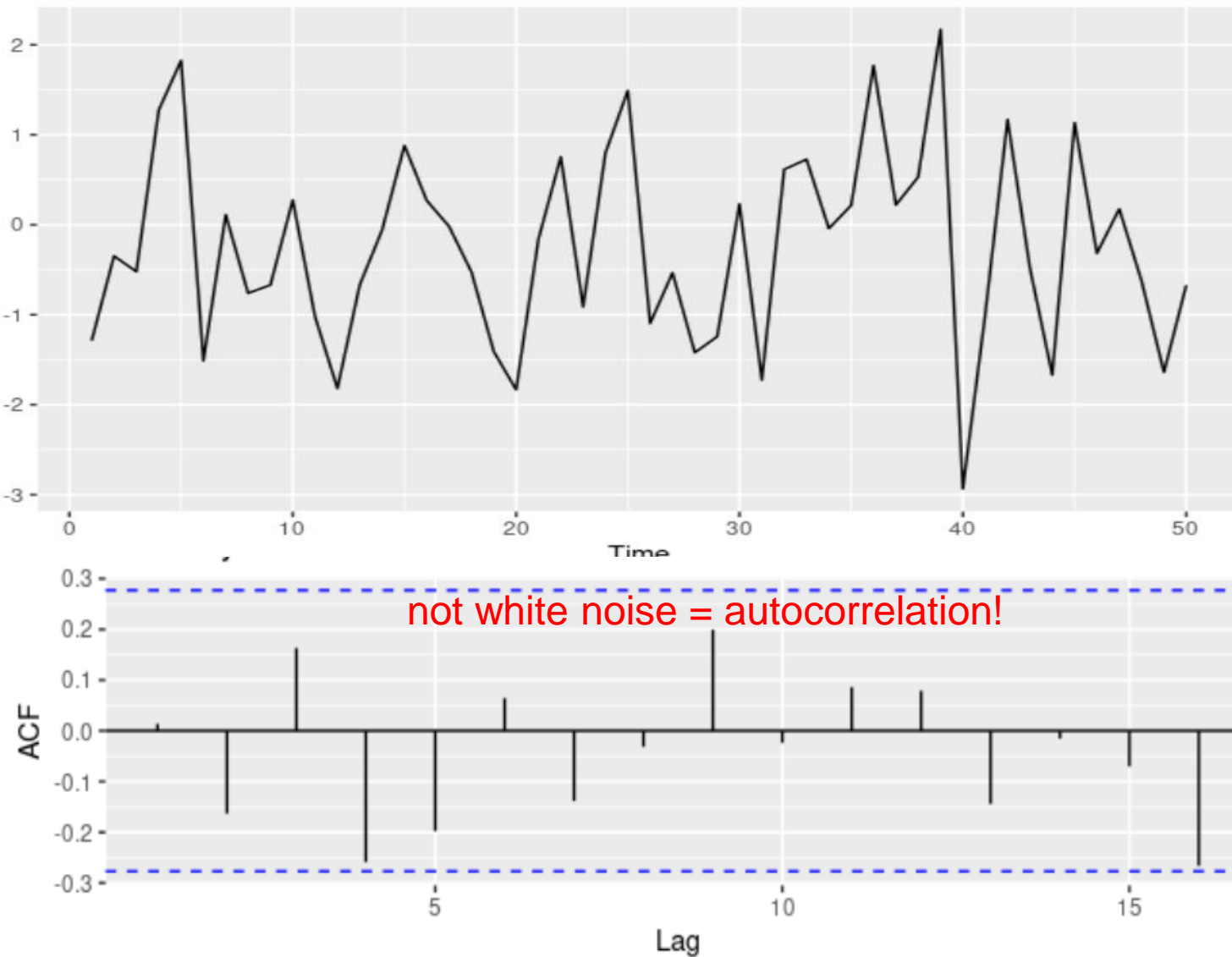
Correlation is a single statistic or data point, whereas regression is the entire equation with all of the data points that are represented with a line.

Correlation shows the relationship between two variables, while regression allows us to see how one affects the other.

Difference between Correlation and Regression

Basis For Comparison	Correlation	Regression
Meaning	Correlation is a statistical measure that determines the association or co-relationship between two variables.	Regression describes how to numerically relate an independent variable to the dependent variable.
Usage	To represent a linear relationship between two variables.	To fit the best line and to estimate one variable based on another.
Dependent and Independent variables	No difference	Both variables are different.
Indicates	Correlation coefficient indicates the extent to which two variables move together.	Regression indicates the impact of a change of unit on the estimated variable (y) in the known variable (x).
Objective	To find a numerical value expressing the relationship between variables.	To estimate values of random variables on the basis of the values of fixed variables.

Time series that show no autocorrelation are called white noise



For white noise series, we expect each autocorrelation to be close to zero. They will not be exactly equal to zero as there is some random variation. We expect 95% of the spikes in the ACF to lie within $\pm 2/\sqrt{T}$ where T is the length of the time series.

It is common to plot these bounds on a graph of the ACF (the blue dashed lines above). If one or more large spikes are outside these bounds, or if substantially more than 5% of spikes are outside these bounds, then the series is probably not white noise.

LSTM (Long Short Term Memory) networks are a special type of RNN (Recurrent Neural Network) that is structured to remember and predict based on long-term dependencies that are trained with time-series data. An LSTM repeating module has some interacting components.

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graph TD; A[Data Generation and Preparation] --> B[LSTM Model Build]; B --> C[LSTM Model Training]; C --> D[LSTM Prediction Validation]; D --> E[LSTM Forecast Validation];
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LSTM Forecast Validation

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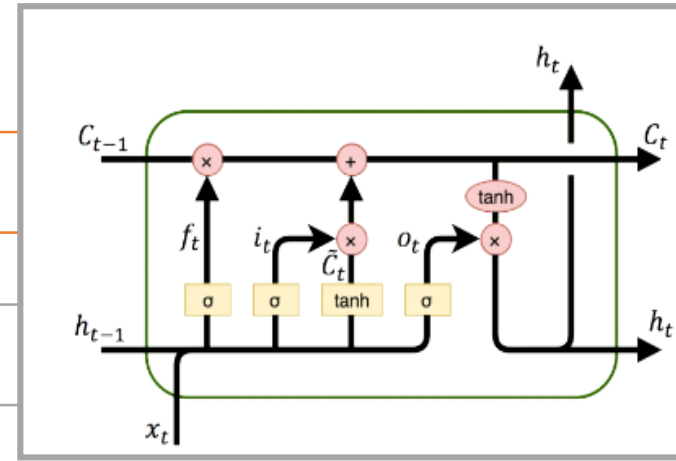
Data Generation and Preparation

LSTM Model Build

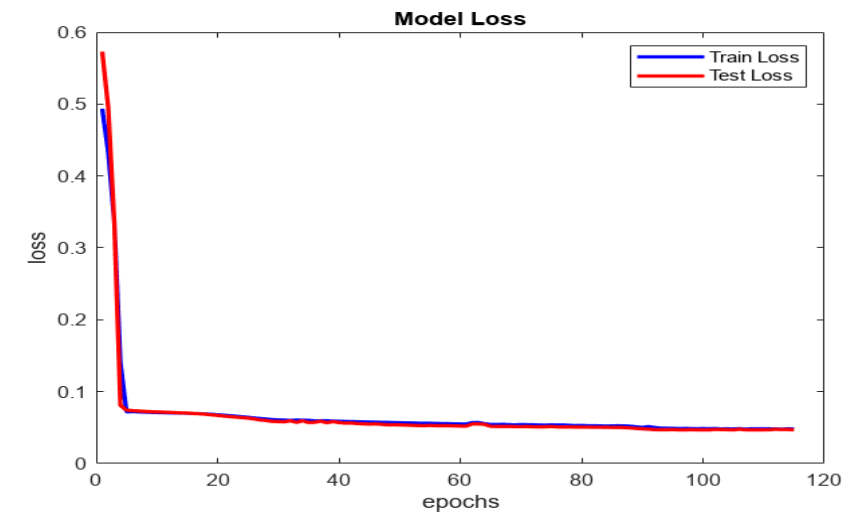
LSTM Model Training

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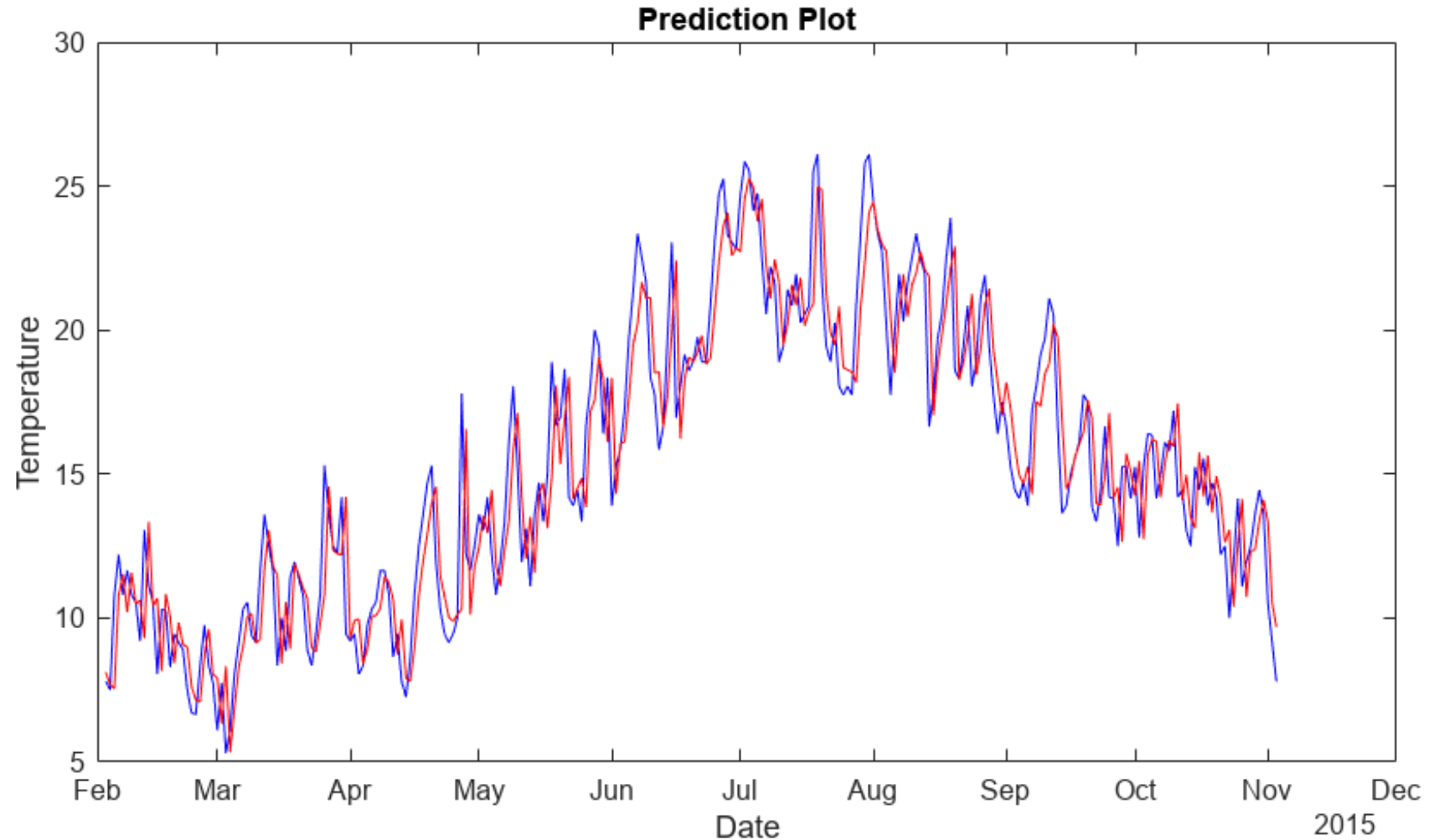


An LSTM network relates the input data window to outputs with layers. Instead of just one layer, LSTMs often have multiple layers.



LSTM Prediction Validation

The validation test set assesses the ability of the neural network to predict based on new conditions that were not part of the training set. The validation is performed with the last 20% of the data that was separated from the beginning 80% of data.



LSTM Forecast Validation

When performing the validation it is also important to determine how the model performs with without measurements when it uses prior predictions to predict the next outcome. This is important to determine how well the model performs in a predictive application such as model predictive control where the model is projected forward over the control horizon to determine the sequence of optimal manipulated variable moves and possible future constraint violation. Generating predictions without measurement feedback is a forecast.

