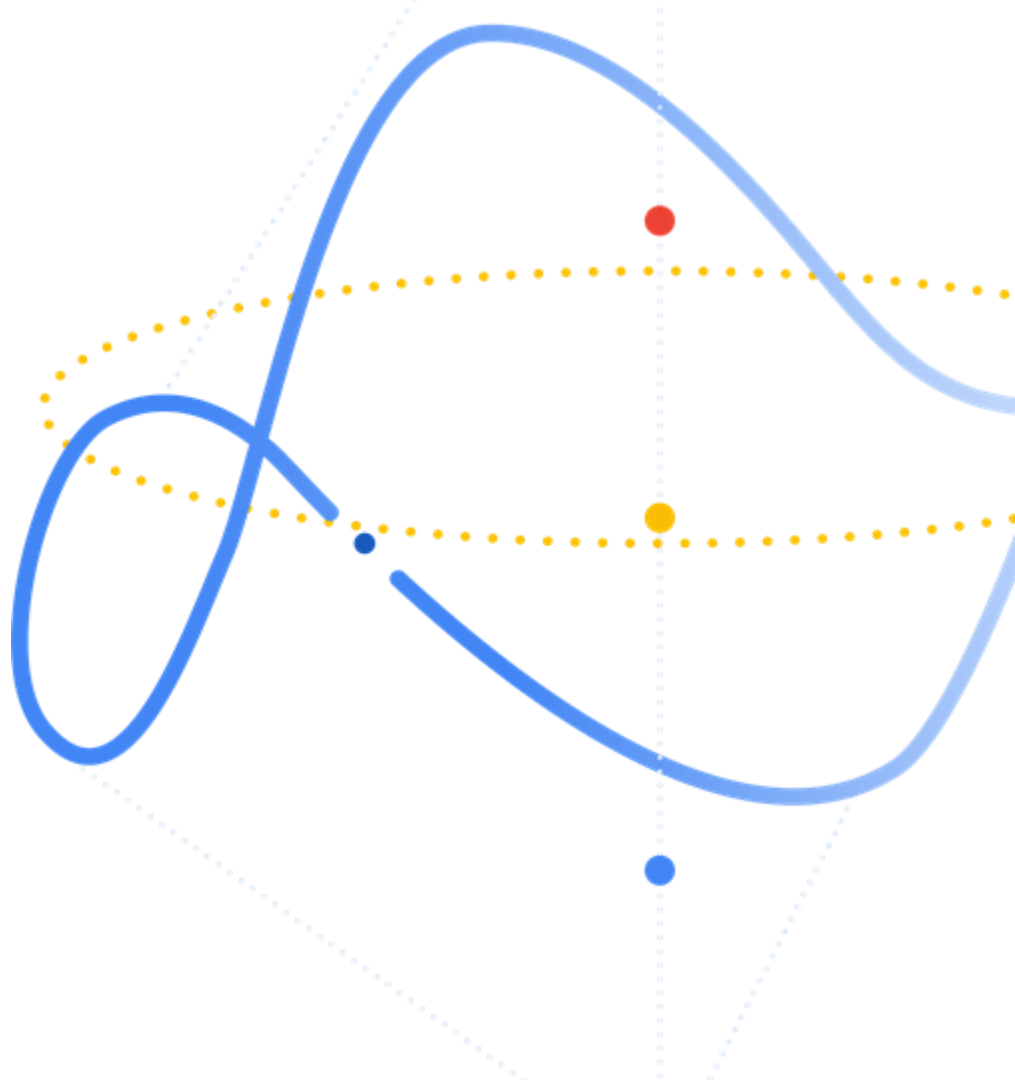


# Machine Learning



# What is Machine Learning?

**Machine Learning systems take inputs (data) to make useful predictions and decisions about previously unseen pieces of data.**

**Machine learning is a specific field of AI where a system learns to find patterns in examples in order to make predictions.**

**Computers learning how to do a task  
without being explicitly programmed to  
do so.**

## Machine Learning systems might:

- Label or classify data
- Predict numerical values
- Cluster similar pieces of data together
- Infer association patterns in data
- Create complex outputs

# "Machine Learning: Why or Why not?"



Read a couple of news articles involving applications of ML.

1. Would a traditional programming solution be more efficient?
1. Could a human perform the same task in less time?
1. What are the benefits of a Machine Learning model in these instances?

# Machine Learning

## Supervised

Model is trained  
on labeled data



stop\_sign\_1



stop\_sign\_2



Google Alstop\_sign\_3



stop\_sign\_4

## Unsupervised

Model learns patterns  
from unlabelled data.





# See it in action!

## Supervised learning



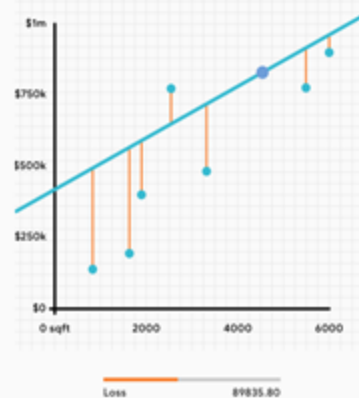
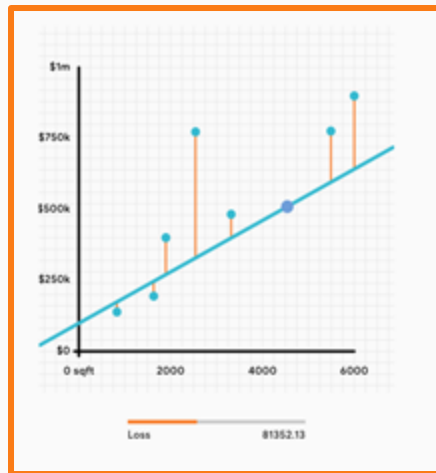
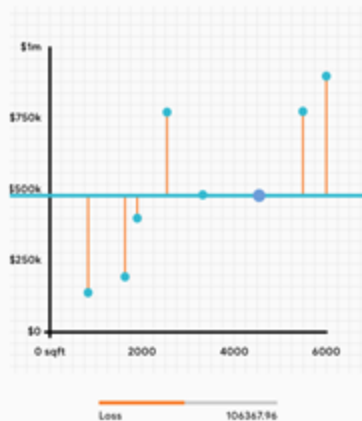
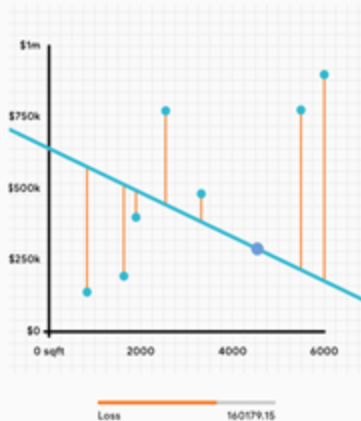
Image label verification

## Unsupervised learning



Semantic Similarity

# Loss



# Supervised

```
graph TD; A[Supervised] --> B[Classification]; A --> C[Regression]; B --> D[Linear Classification]; C --> E[Linear Regression]; B --- F["Predict a discrete class label for a given input"]; C --- G["Predict a continuous numerical value or a quantity"];
```

## Classification

**Predict a discrete class label for a given input**

## Linear Classification

## Regression

**Predict a continuous numerical value or a quantity**

## Linear Regression

# Classification

Categorizing data into predefined classes or categories based on their features.

## Process of Classification:

- Data preparation: Collect and label dataset.
- Feature extraction: Extract relevant features.
- Training: Train classification algorithm.
- Evaluation: Assess model performance.
- Prediction: Predict class labels for new instances.





## Linear Regression

- Predicts continuous values.
- Uses linear decision boundary.
- Assumes linear relationship.
- Simple and interpretable.
- Limited for binary classification tasks.

## K-Nearest Neighbors(KNN)

- Proximity-based classification.
- Nonlinear decision boundaries.
- Requires optimal k selection.
- Simple and interpretable.
- Effective for small to medium-sized datasets.

## Random Forest

- Ensemble of decision trees.
- Reduces overfitting.
- Handles high-dimensional data.
- Provides feature importance.
- Robust to missing data and outliers.

# Regression

A statistical technique used to predict continuous numerical values based on the relationship between a dependent variable and one or more independent variables.

## Process of Regression:

- Data preparation: Collect and preprocess dataset.
- Model selection: Choose regression algorithm.
- Training: Fit model to training data.
- Evaluation: Assess model performance.
- Prediction: Make predictions on new data instances.





## Logistic Regression

- Predicts binary outcomes.
- Estimates probabilities.
- Assumes linear relationship.
- Interpretable coefficients.
- Requires feature scaling.

## K-Nearest Neighbors(KNN)

- Uses nearby neighbors for prediction.
- Handles nonlinearity.
- Sensitive to choice of neighbors (k).
- Simple and interpretable.
- Good for small to medium-sized datasets.

## Random Forest

- Ensemble of decision trees.
- Reduces overfitting.
- Handles high-dimensional data.
- Provides feature importance.
- Robust to missing data and outliers.