

Mastering Machine Learning Model Training

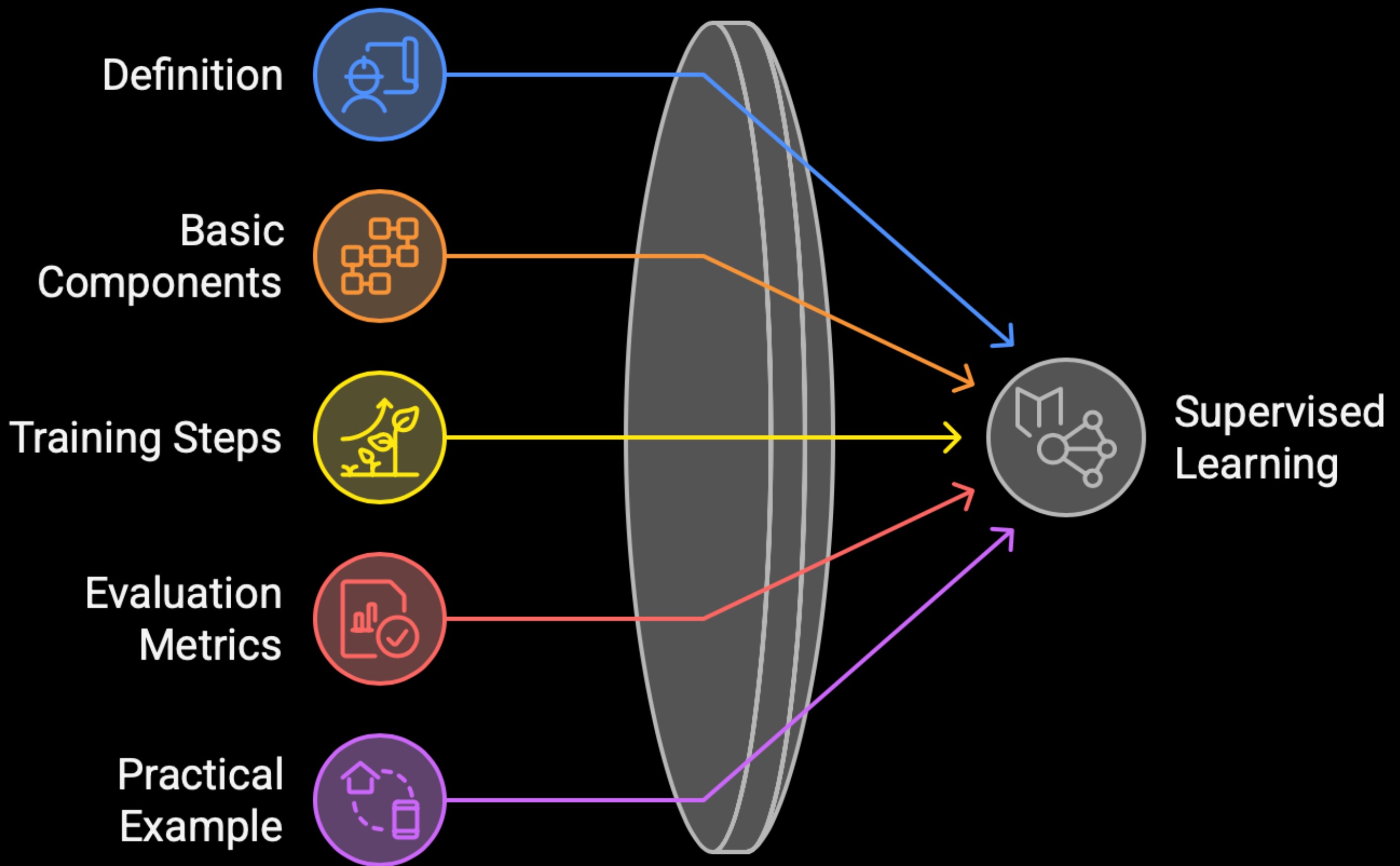
A Complete Guide to Building Supervised Models



Edgar Rios Linares

Roadmap

Mastery of supervised learning



Definition

We "teach" a machine from labeled data so that it learns to predict future outcomes.



Definition

Supervised learning is a technique in which a model is trained with a set of labeled data.

Each training example is accompanied by a desired output.

The goal is for the model to learn to map inputs (features) to correct outputs (labels).

Components

Labeled dataset

Features

Labels

Objective function



Components

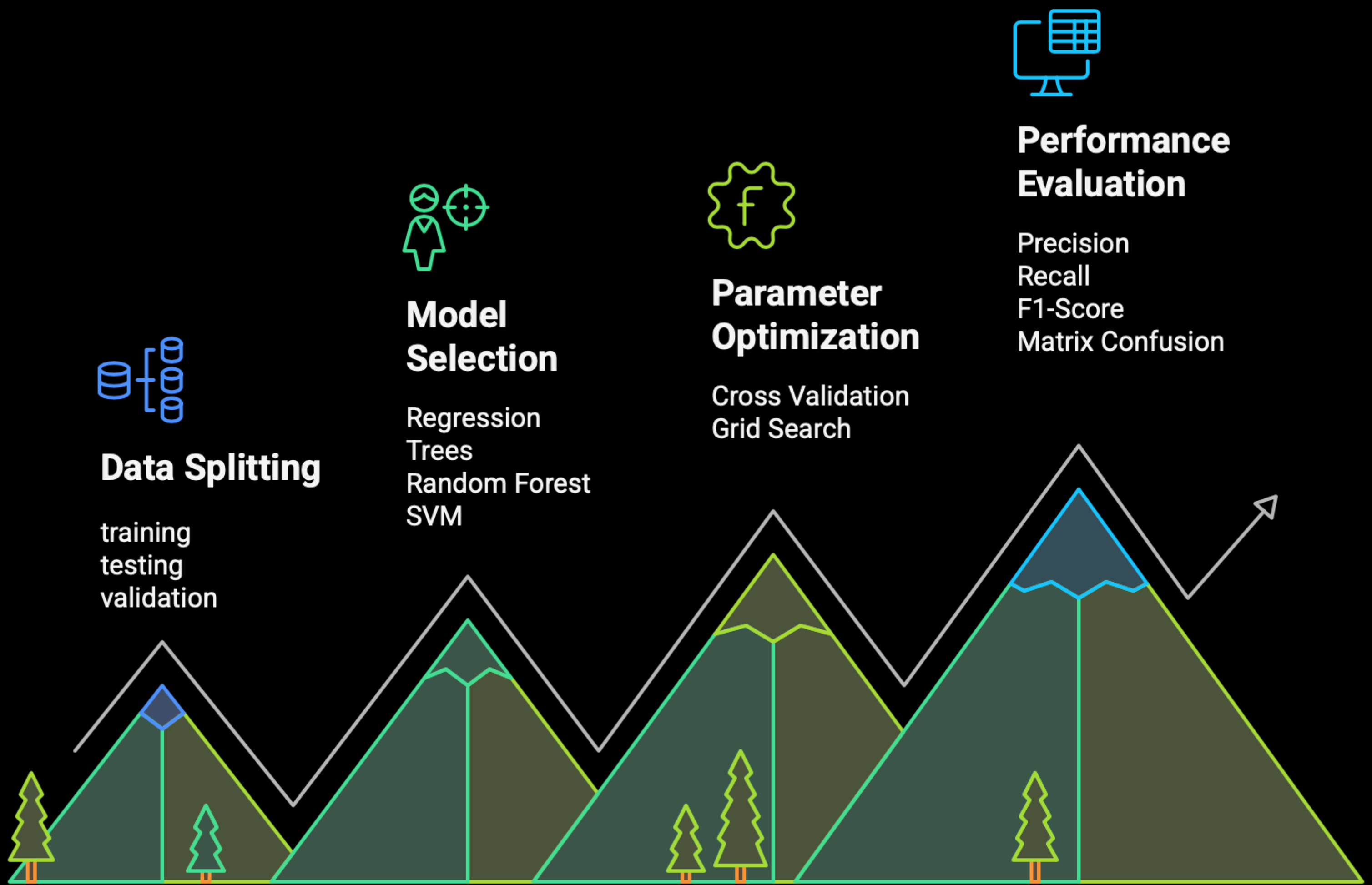
Labeled dataset: Input data with its respective expected outputs.

Features: Relevant attributes that will be used to make predictions (pixels of an image)

Labels: Target values that the model must learn to predict (dog or cat)

Objective Function: Metric that the model optimizes during training to improve its predictions.

Training Steps



Training Steps

1. Data Splitting

Training: Train the model. (70-80%)

Validation: Tune hyperparameters. (10-15%)

Testing: Evaluate final performance. (10-15%)

2. Model Selection

Regression, trees, SVM, networks, etc.

3. Parameter Optimization

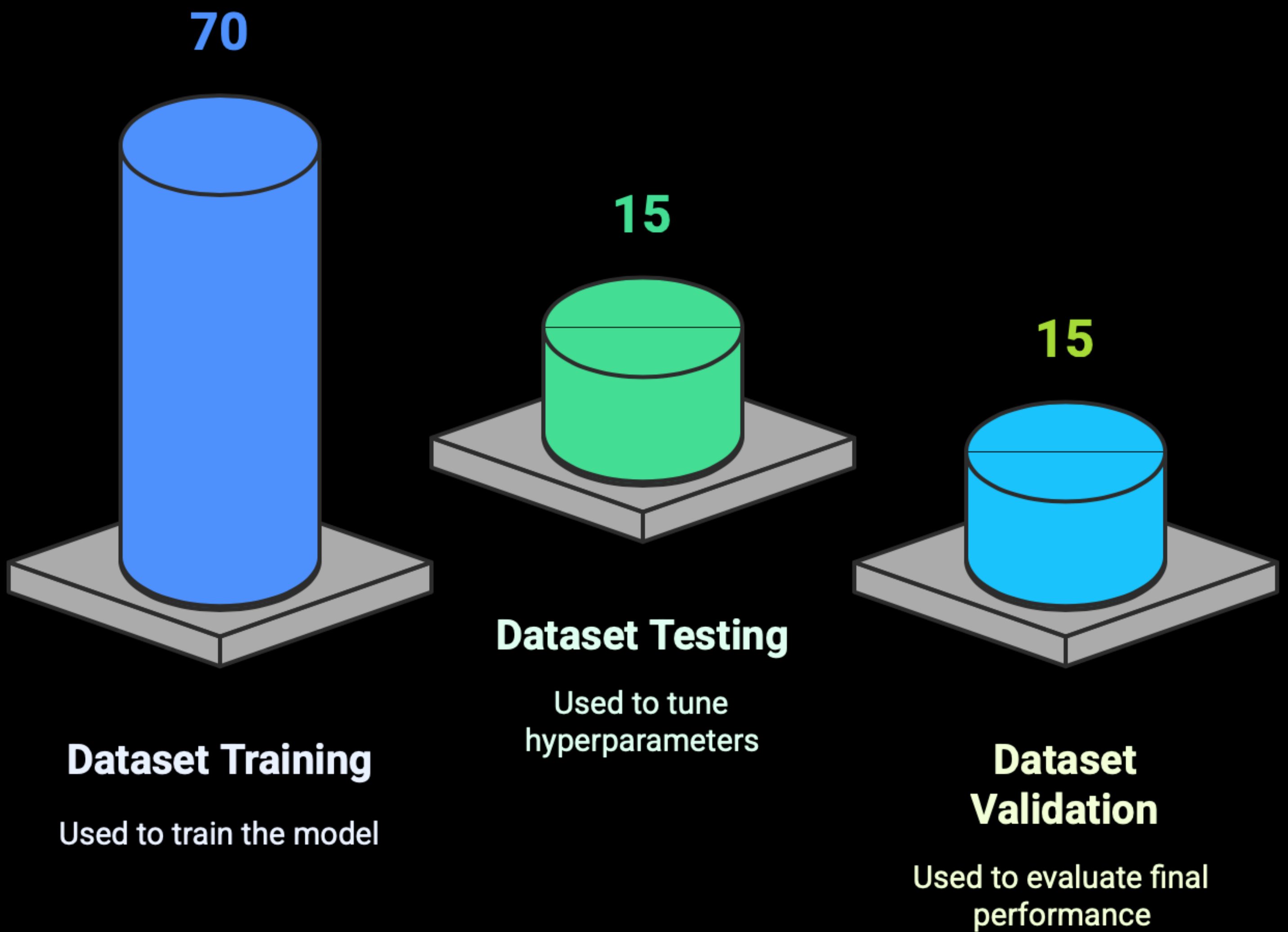
Cross-validation, Grid Search, etc.

4. Performance Evaluation

Validation and testing performance.

Training Steps

1. Data Splitting



Training Steps

1. Data Splitting

Features (X) Label (y)

X1	X2	...	Xn	y	
999	787	678	554	CAT	Training 70%
123	332	444	555	DOG	
123	333	405	554	DOG	
999	787	678	554	CAT	
123	333	405	554	DOG	
999	787	678	554	CAT	
999	787	678	554	CAT	
123	787	405	554	DOG	Test 15%
999	787	678	554	CAT	
999	787	678	554	CAT	Validation 15%
123	333	543	233	DOG	

Training Steps

2. Model Selection



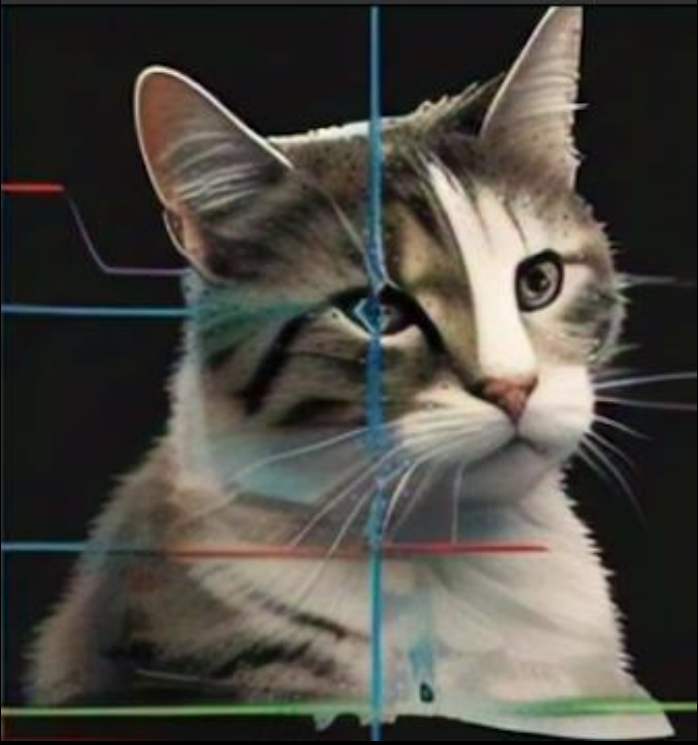
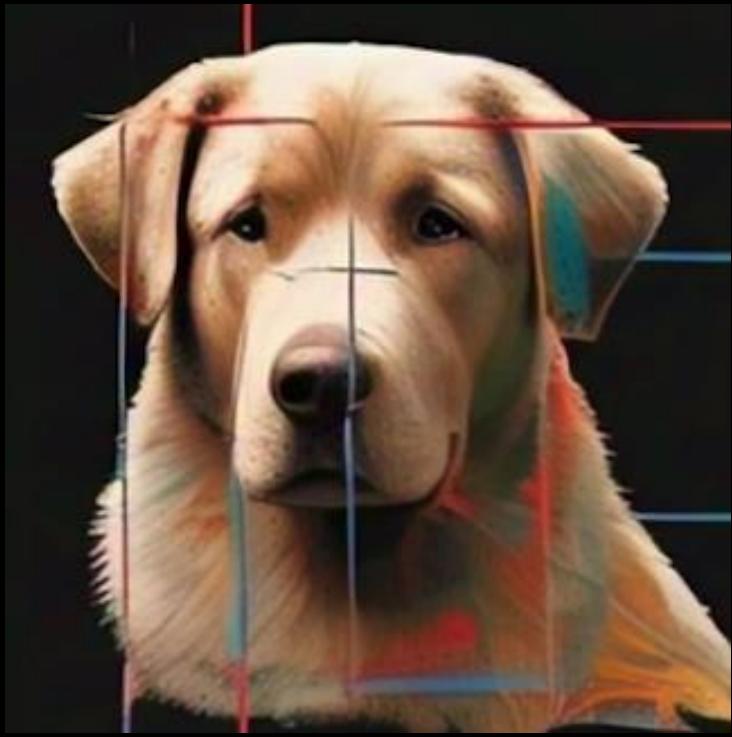
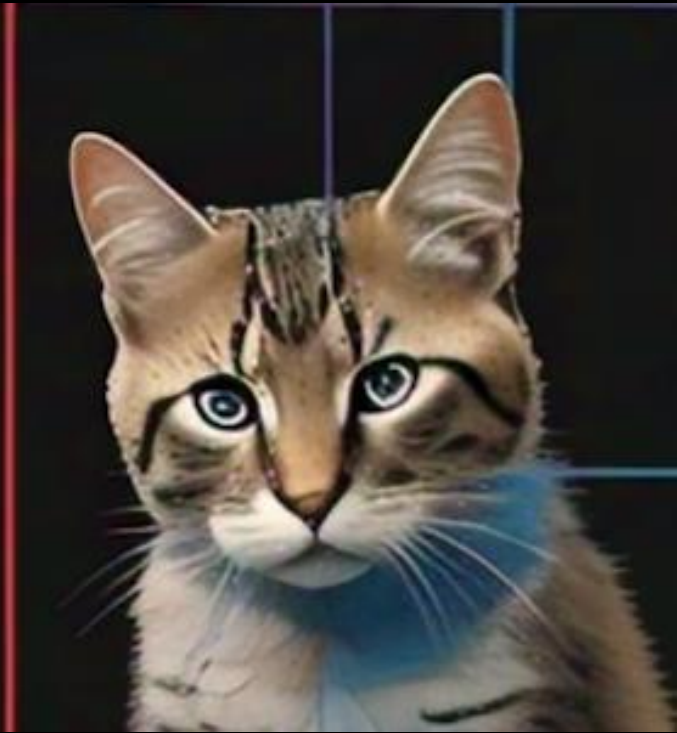
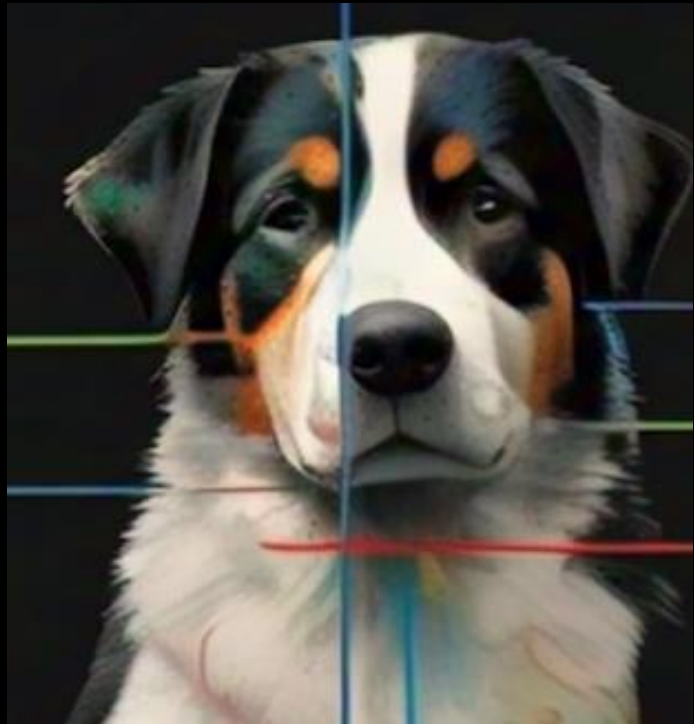
Training Steps

3. Parameter Optimization

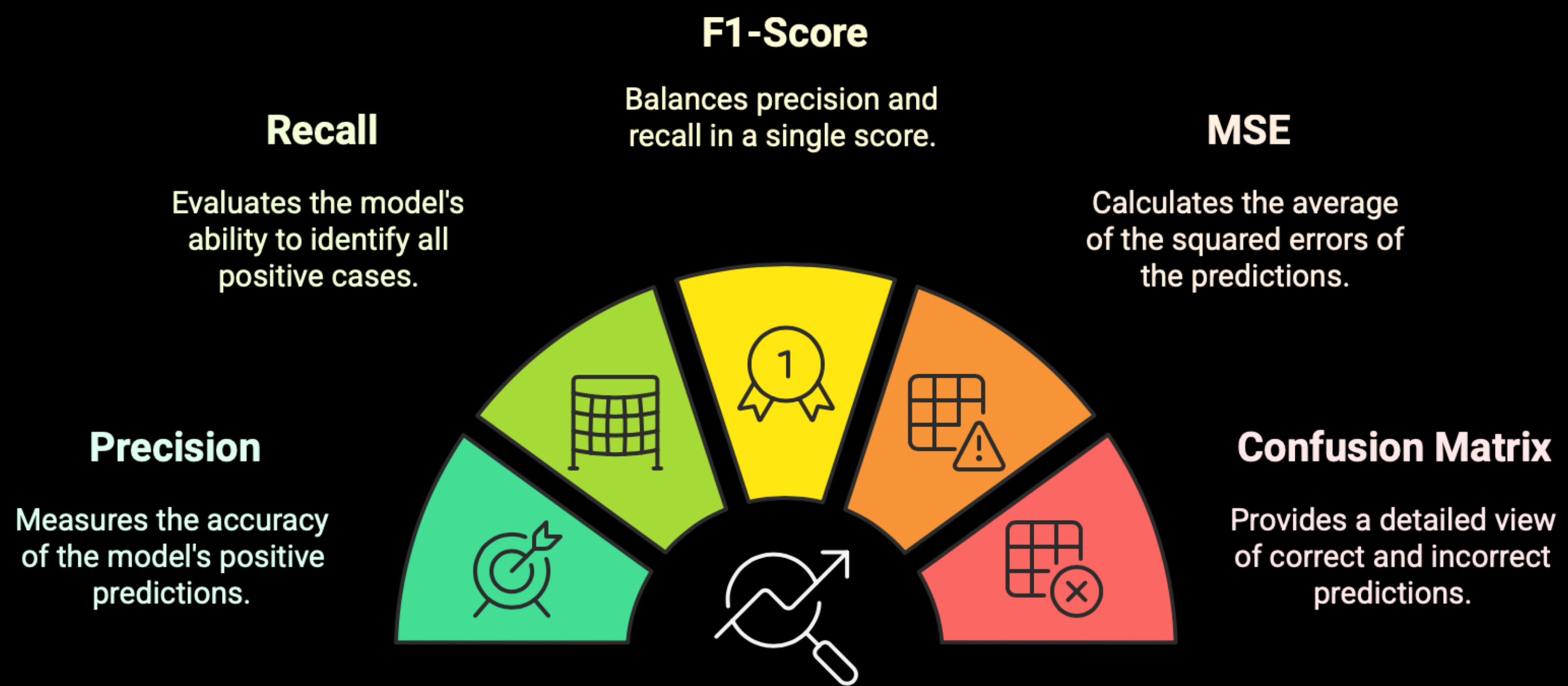
Features (X)				Label (y)
X1	X2	...	Xn	y
999	787	678	554	CAT
123	332	444	555	DOG
123	333	405	554	DOG
999	787	678	554	CAT
123	333	405	554	DOG
999	787	678	554	CAT
999	787	678	554	CAT
123	787	405	554	DOG
999	787	678	554	CAT
999	787	678	554	CAT
123	333	543	233	DOG

Training Steps

4. Performance Evaluation

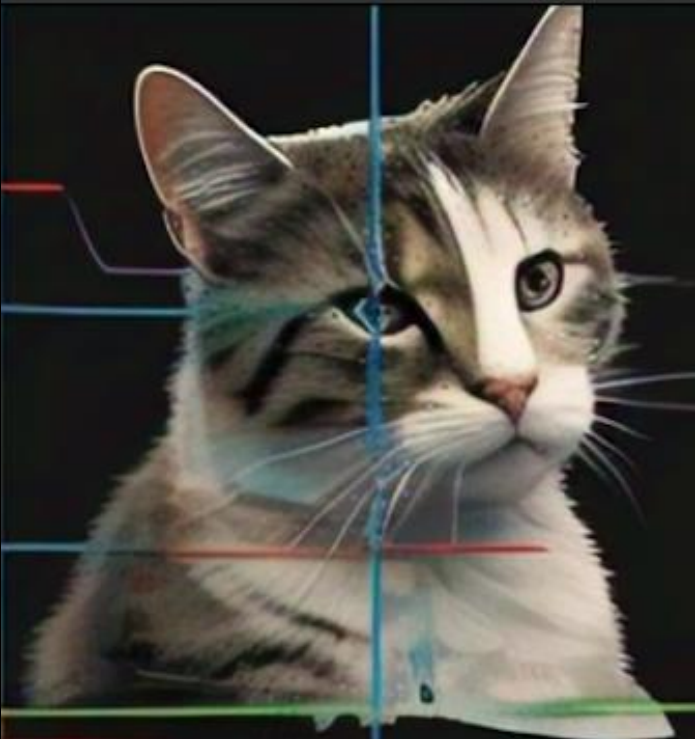
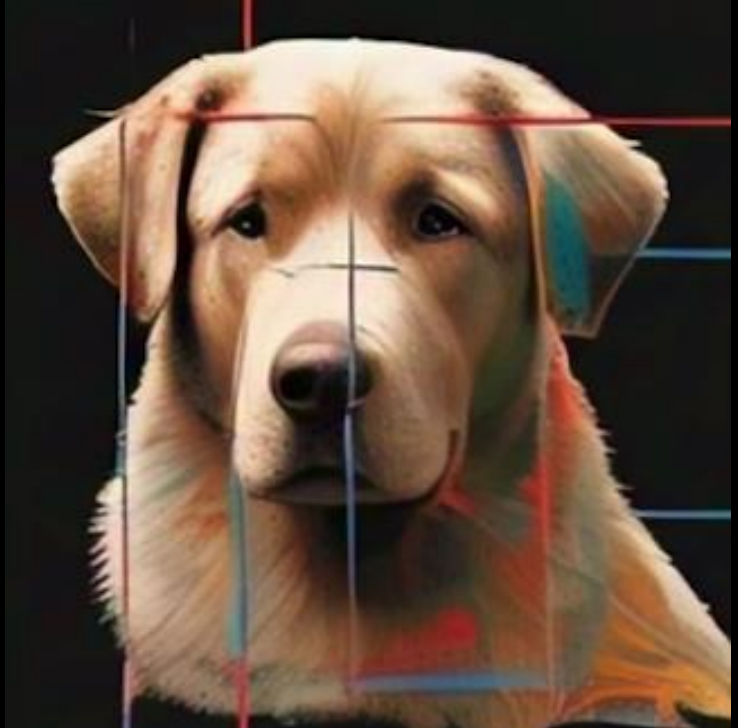
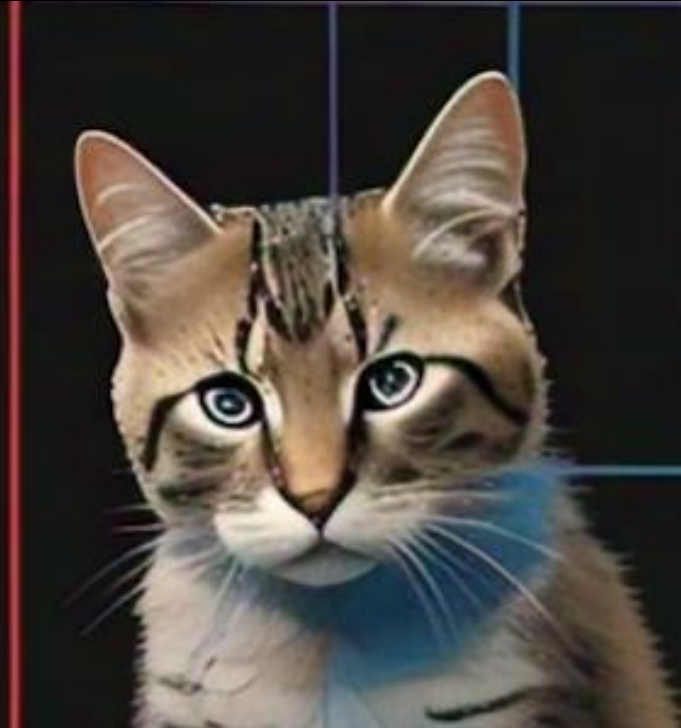
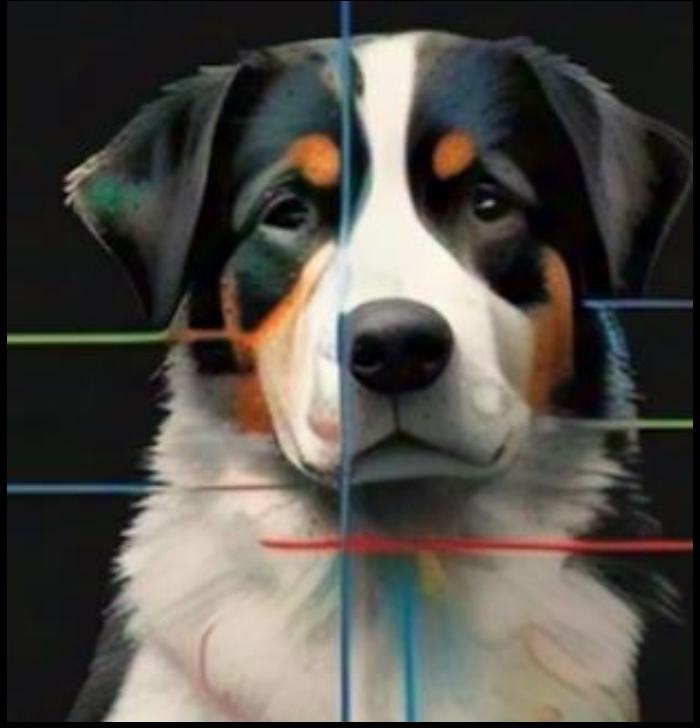
		ACTUAL	
		CAT	NO CAT
PREDICTION	CAT	 TRUE POSITIVE	 FALSE POSITIVE ERROR
	NO CAT	 FALSE NEGATIVE ERROR	 TRUE NEGATIVE

Evaluation Metrics



Evaluation Metrics

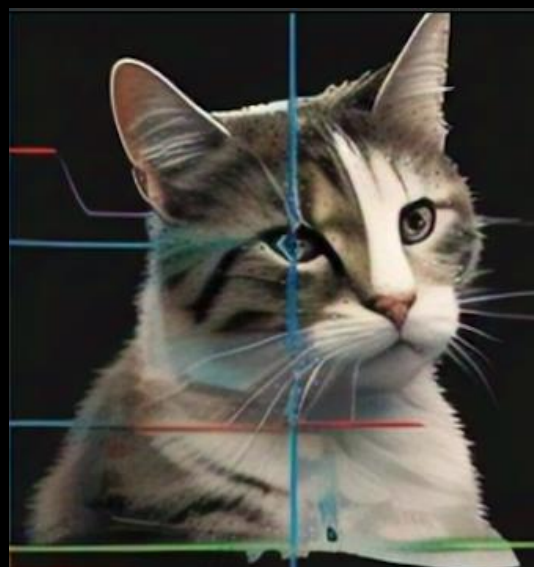
Confusion Matrix, It shows the number of correct and incorrect predictions by comparing the actual labels with the model predictions

		ACTUAL	
		CAT	NO CAT
PREDICTION	CAT	 TRUE POSITIVE	 FALSE POSITIVE ERROR
	NO CAT	 FALSE NEGATIVE ERROR	 TRUE NEGATIVE

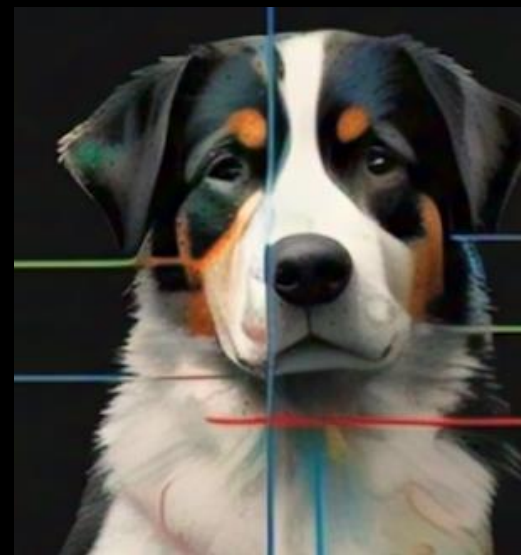
Evaluation Metrics

Precision (accuracy) Proportion of correct predictions out of the total number of predictions.

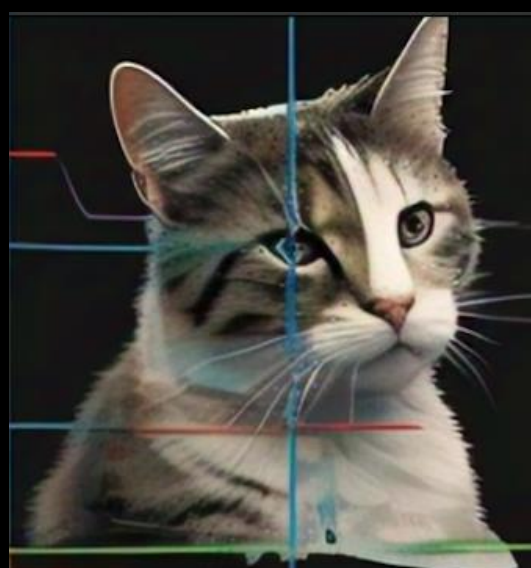
Formula: (True positives + True negatives) / Total.



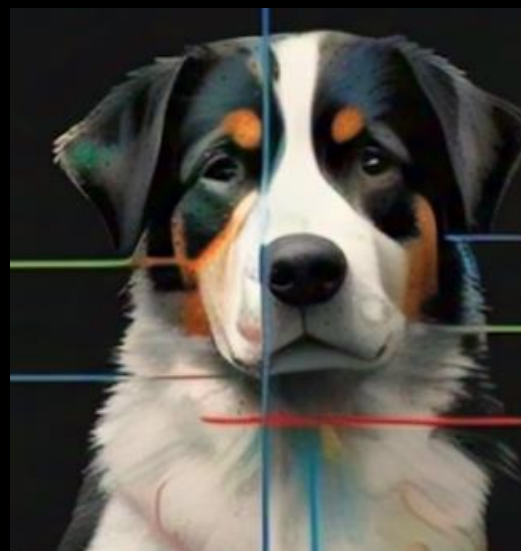
TRUE POSITIVE



TRUE NEGATIVE



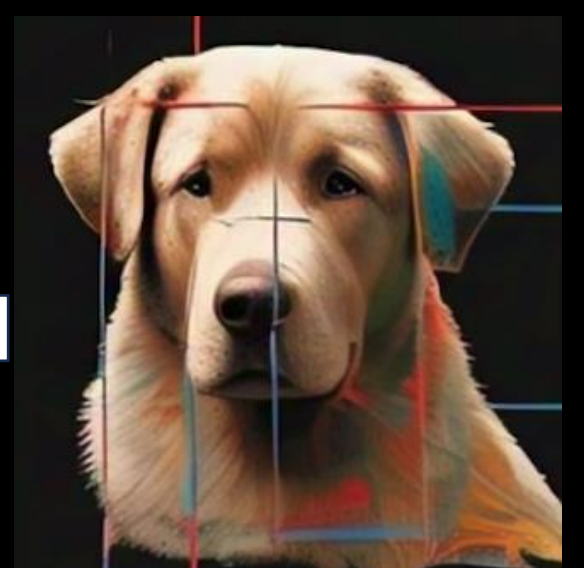
TRUE POSITIVE



TRUE NEGATIVE



FALSE NEGATIVE

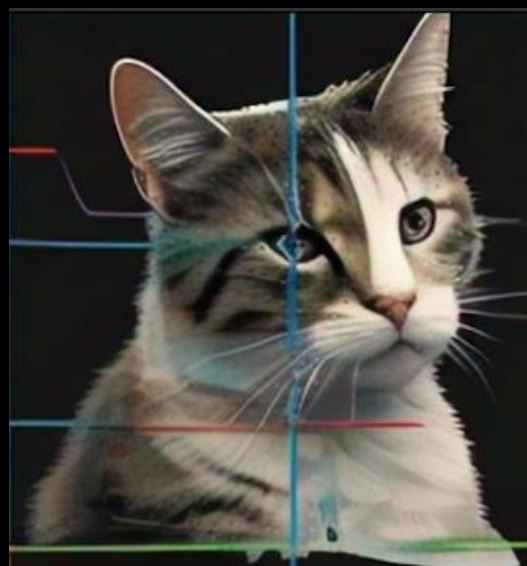


FALSE POSITIVE

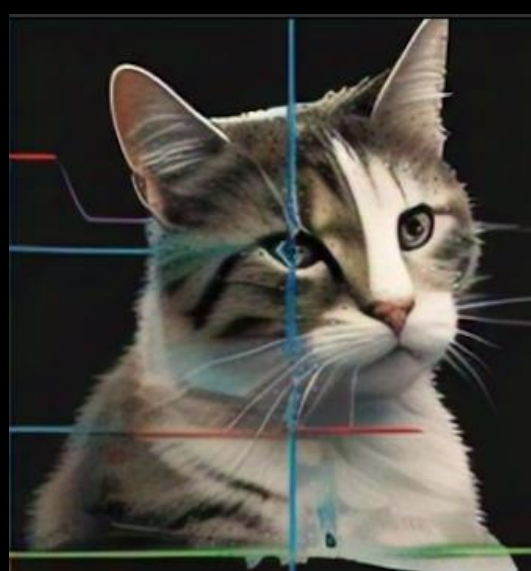
Evaluation Metrics

Recall (Sensitivity) Proportion of true positives out of the total number of real positives.

Formula: $\text{True positives} / (\text{True positives} + \text{False negatives})$.



TRUE POSITIVE



TRUE POSITIVE



FALSE NEGATIVE

Evaluation Metrics

F1-Score, Average of precision and recall. Useful when there is imbalance in classes.

Formula: $2 * (\text{Precision} * \text{Recall}) / (\text{Precision} + \text{Recall})$.

$$F1\text{-score} = 2 \times \frac{\text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

Evaluation Metrics

MSE (Mean Square Error): For regression problems, measures the difference between actual and predicted values

$$MSE = \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

Where :

y_i are the actual values.

\hat{y}_i are the predictions of the model.

n is the total number of observations.

Practical Example

Predicting Solvent Companies

Step 1: Data Collection

A data set is obtained with characteristics such as income, expenses, credits and solvency.

Step 2: Preprocessing

Data cleaning and treatment of null values.

Normalization of numerical variables.

Step 3: Model Training

Data division into training, validation and testing.

LDA model - relationship of characteristics and solvency.

Step 4: Evaluation

Precision, Recall, F1-Score and Confusion Matrix are calculated.

Hyperparameters are adjusted and retraining is carried out.

Practical Example

STEP 1

	income	debt	credit_history	solvent
0	57450.712295	26483.242833	392.115420	0
1	47926.035482	33365.916483	357.512312	0
2	59715.328072	10210.026983	650.036637	1
3	72845.447846	23940.784657	688.561650	1
4	46487.699379	15445.502016	317.372380	1

STEP 2

	income	debt	credit_history
0	-0.245825	-0.531158	1.500187
1	-0.042386	0.079393	-0.803370
2	-0.741307	-0.848214	-0.136724
3	-1.040191	0.480125	0.156964
4	-0.720783	0.168569	-1.356440

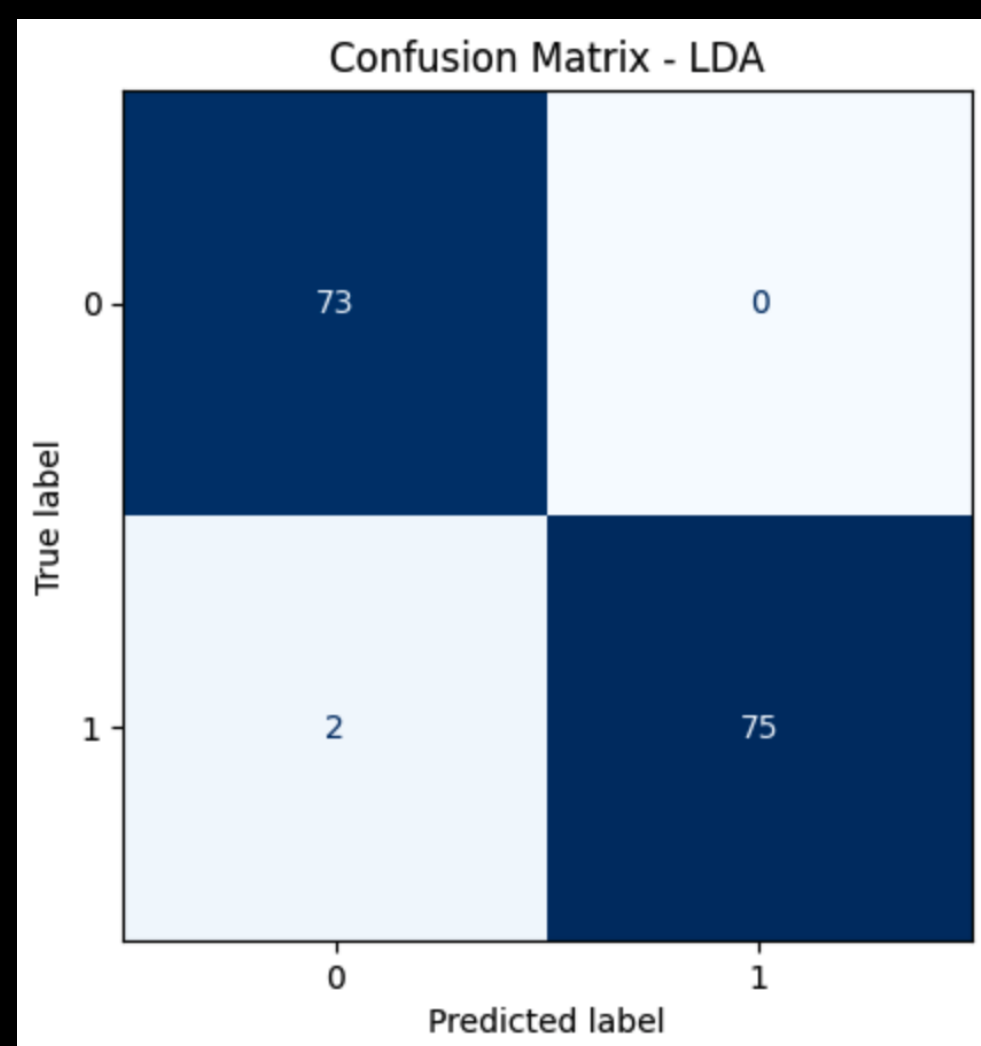
STEP 3

```
X_lda
```

▼ LinearDiscriminantAnalysis ⓘ ?

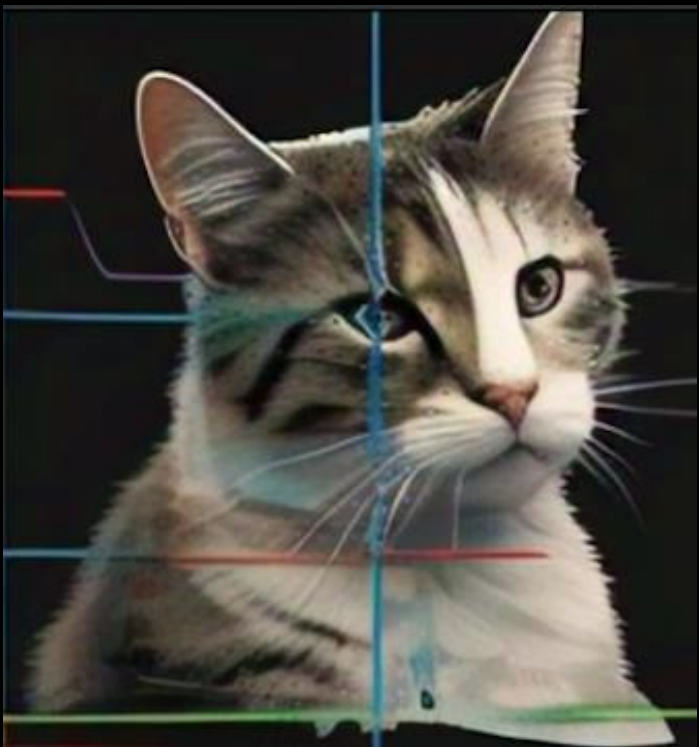
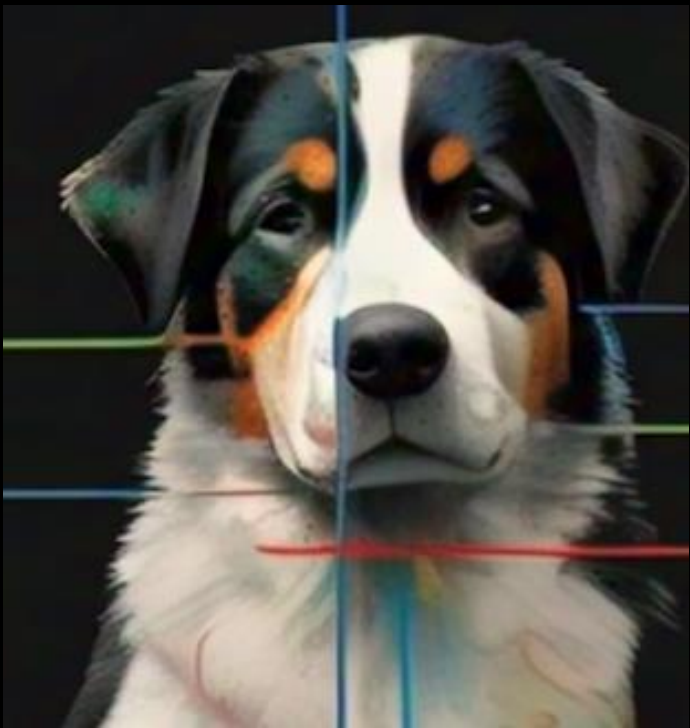
LinearDiscriminantAnalysis()

STEP 4



OVERFITTING

Overfitting occurs when a model learns the details and noise of the training set too well, but does not generalize well to new data. It memorizes the data.

		ACTUAL	
		CAT	NO CAT
PREDICTION	CAT	 TRUE POSITIVE	0
	NO CAT	0	 TRUE NEGATIVE

Educator in AI

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Intelligence**

Data Engineering



Machine Learning

Data Science

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