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# FRAMING: KEY ML TERMINOLOGY

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What's (supervised) machine learning? Concisely put, it's the following:

- ML systems learn how to combine input to produce useful predictions on never-before-seen data

Let's explore fundamental machine learning terminology.

## Labels

A **label** is the thing we're predicting—the  $y$  variable in simple linear regression. The label could be the future price of wheat, the kind of animal show in a picture, the meaning of an audio clip, or just about anything.

## Features

A **feature** is an input variable—the  $x$  variable in simple linear regression. A simple machine learning project might use a single feature, while a more sophisticated machine learning project could use millions of features, specified as:

$$x_1, x_2, \dots, x_n$$

In the spam detector example, the features could include the following:

- Words in the email text
- Sender's address
- Time of day email was sent
- Email contains the phrase "one weird trick"

## Examples

An **example** is a particular instance of data,  $\mathbf{x}$  (We put  $\mathbf{x}$  in boldface to indicate that it's a vector.) We break examples into two categories:

- Labeled examples
- Unlabeled examples

A **labeled example** includes both feature(s) and the label, That's:

```
> Labeled examples: {features, label}: (x, y)
```

Use labeled examples to **train** the model. In our spam detector example, the labeled examples would be individual emails that users have explicitly marked as “spam” or “not spam”.

For example, the following table shows 5 labeled examples from a data set containing information about housing prices in California:

housingMedianAge (feature)	totalRooms (feature)	totalBedrooms (feature)	medianHouseValue (label)
15	5612	1283	66900
19	7650	1901	80100
17	720	174	85700
14	1501	337	73400
20	1454	326	65500

An unlabeled example contains features but not the label. That’s:

```
> unlabeled examples: {features, ?}: (x, ?)
```

Here are 3 unlabeled examples from the same housing dataset, which exclude `medianHouseValue`:

housingMedianAge (feature)	totalRooms (feature)	totalBedrooms (feature)
42	168	361
34	1226	180
33	1077	271

Once we've trained our model with labeled examples, we use that model to predict the label on unlabeled examples. In the spam detector, unlabeled examples are new emails that humans haven't yet labeled.

## Models

A model defines the relationship between features and label. For example, a spam detection model might associate certain features strongly with "spam". Let's highlight two phases of a model's life:

- **Training** means creating or **learning** the model. That is, you show the model labeled examples and enable the model to gradually learn the relationships between features and label.
- **Inference** means applying the trained model to unlabeled examples. That is, you use the trained model to make useful predictions (y'). For example, during inference, you can predict `medianHouseValue` for new unlabeled examples.

## Regression vs. classification

A **regression** model predicts continuous values. For example, regression models make predictions that answer questions like the following:

- What is the value of a house in California?
- What is the probability that a user will click on this ad?

A **classification** model predicts discrete values. For example, classification models make predictions that answer questions like the following:

- Is a given email message spam or not spam?
- Is this an image of a dog, a cat, or a hamster?