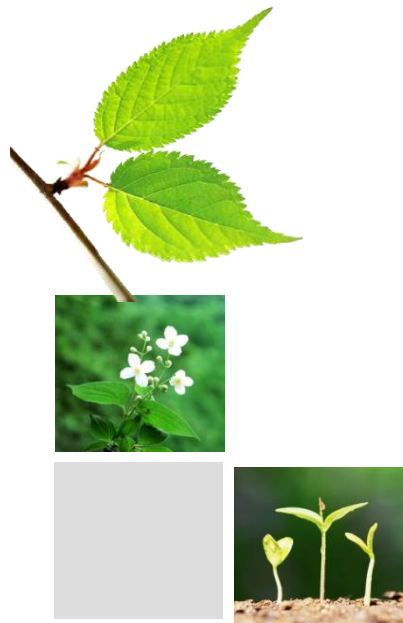


CS231. Nhập môn Thị giác máy tính

Machine Learning Image Classification: Linear classifiers



Mai Tiến Dũng

Targets :



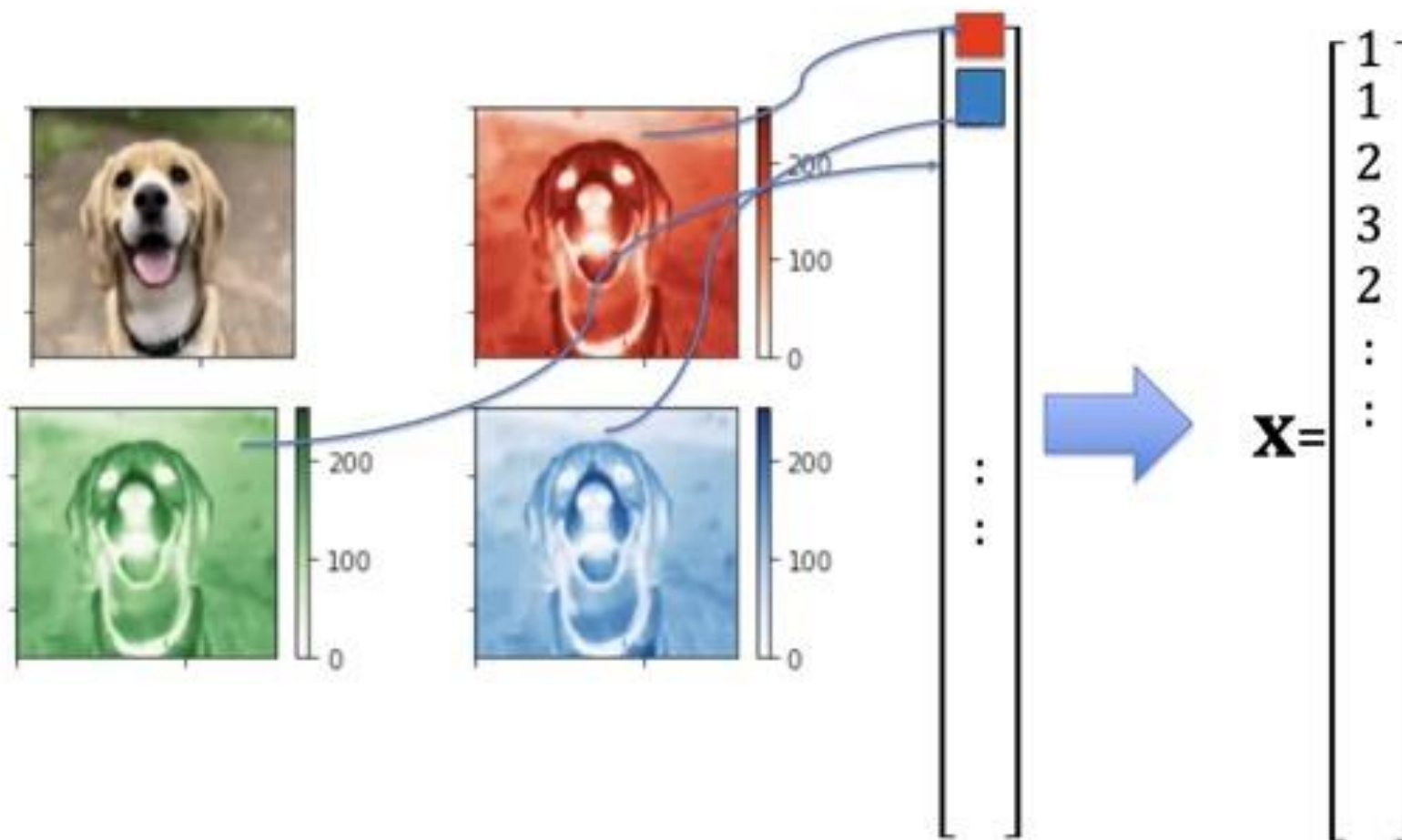
Targets :

$y = 0$




$y = 1$






Function:

$$\hat{y} = f(\text{)$$

The diagram illustrates a function f that takes an input x and produces an output \hat{y} . The input x is represented by a photograph of a happy beagle dog with its tongue out. A bracket above the photo is labeled with the variable x , indicating that the photo is the input to the function.

Function:

$$\hat{y} = f(\text{) = 1$$

The diagram illustrates a function f that takes an input x and produces an output \hat{y} . The input x is represented by a photograph of a beagle dog, which is labeled with a bracket and the variable x . The output \hat{y} is shown to be 1.

Function:

Probability this image is a dog

$$p(\text{Image of a dog}) = 0.92$$

Linear Classifier

- Learnable Parameters

$$z = wX + b$$

Linear Classifier

- Learnable Parameters

$$z = wX + b$$

$$z = w_1x_1 + \dots + w_dx_d + b$$

Linear Classifier

- Learnable Parameters

$$z = \mathbf{w}\mathbf{x} + b$$

$$z = w_1x_1 + \dots + w_dx_d + b$$

$$z = \mathbf{w}\mathbf{x} + b$$

Linear Classifier

$$z = \mathbf{w}\mathbf{x} + \mathbf{b}$$

$$z = w_1x_1 + w_2x_2 + \mathbf{b}$$

Linear Classifier

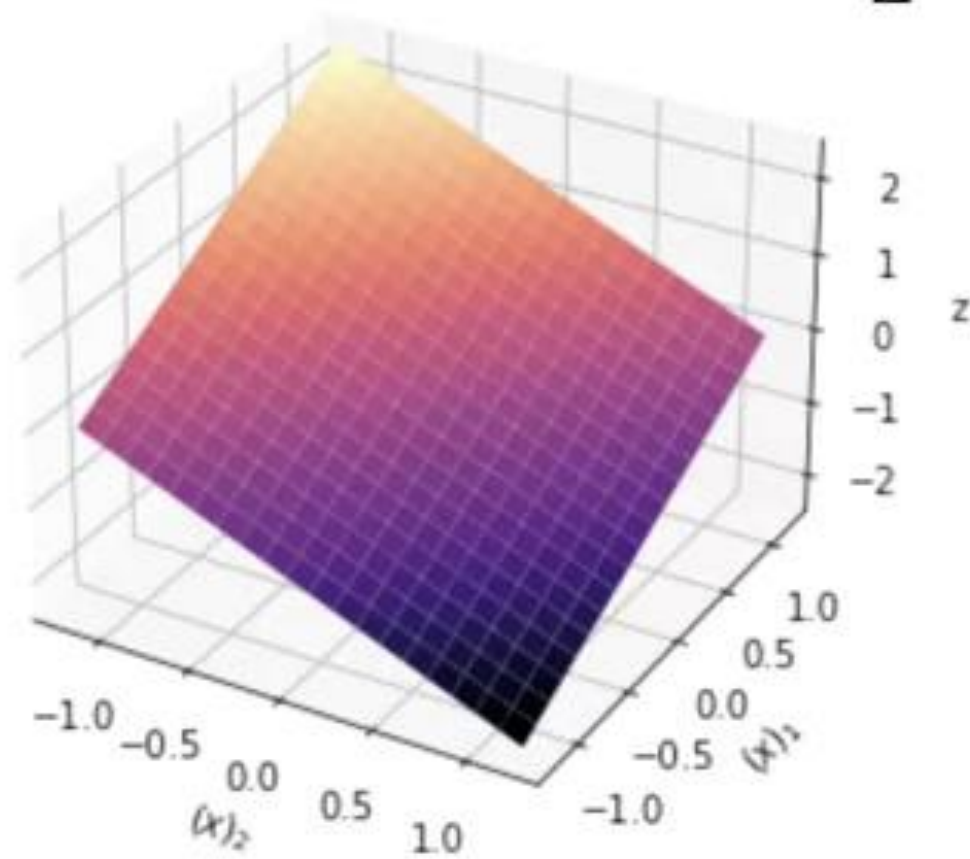
$$z = \mathbf{w}\mathbf{x} + \mathbf{b}$$

$$z = w_1x_1 + w_2x_2 + \mathbf{b}$$

$$z = 1x_1 - 1x_2 + 1$$

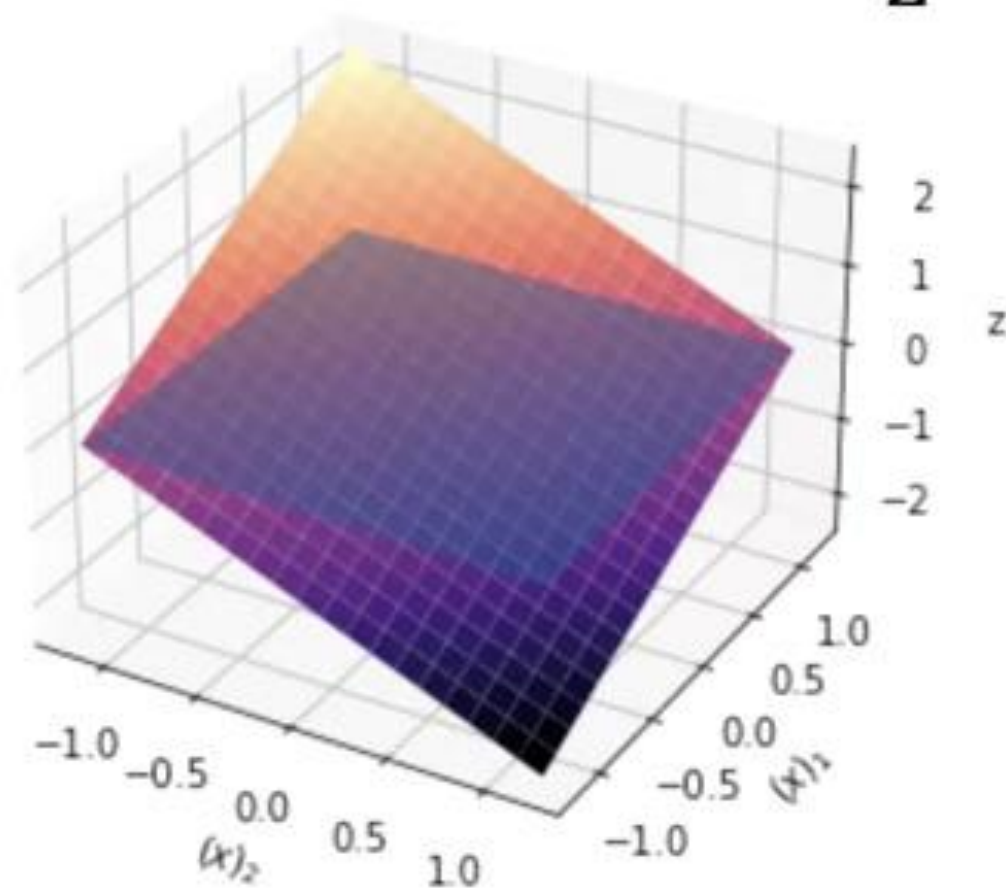
$$z = w_1x_1 + w_2x_2 + b$$

$$z = \mathbf{w}\mathbf{x} + b$$



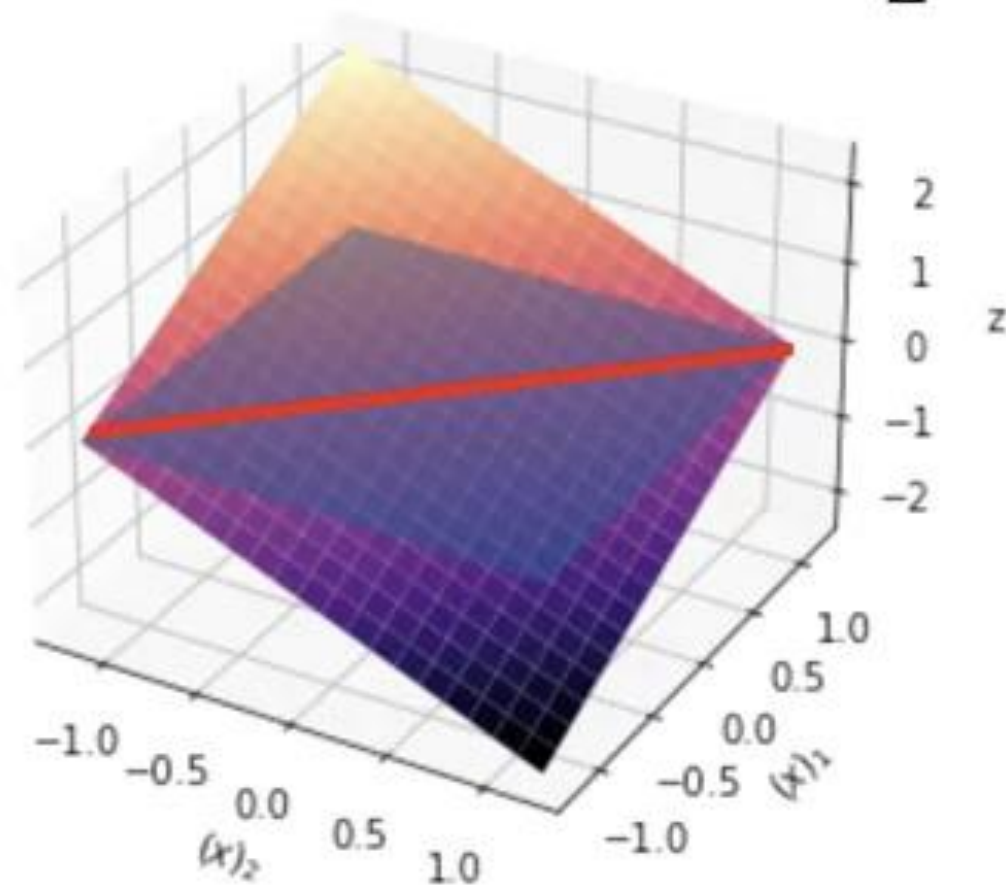
$$z = w_1x_1 + w_2x_2 + b$$

$$z = \mathbf{w}\mathbf{x} + b$$



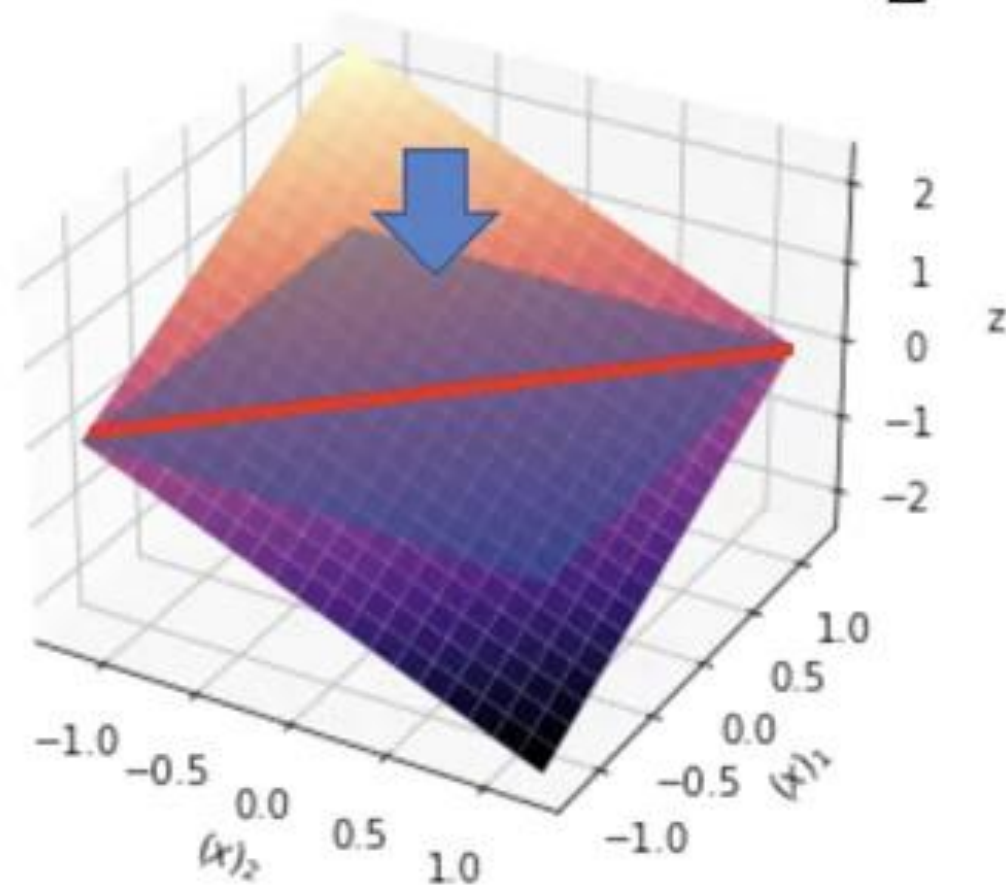
$$z = w_1x_1 + w_2x_2 + b$$

$$z = \mathbf{w}\mathbf{x} + b$$

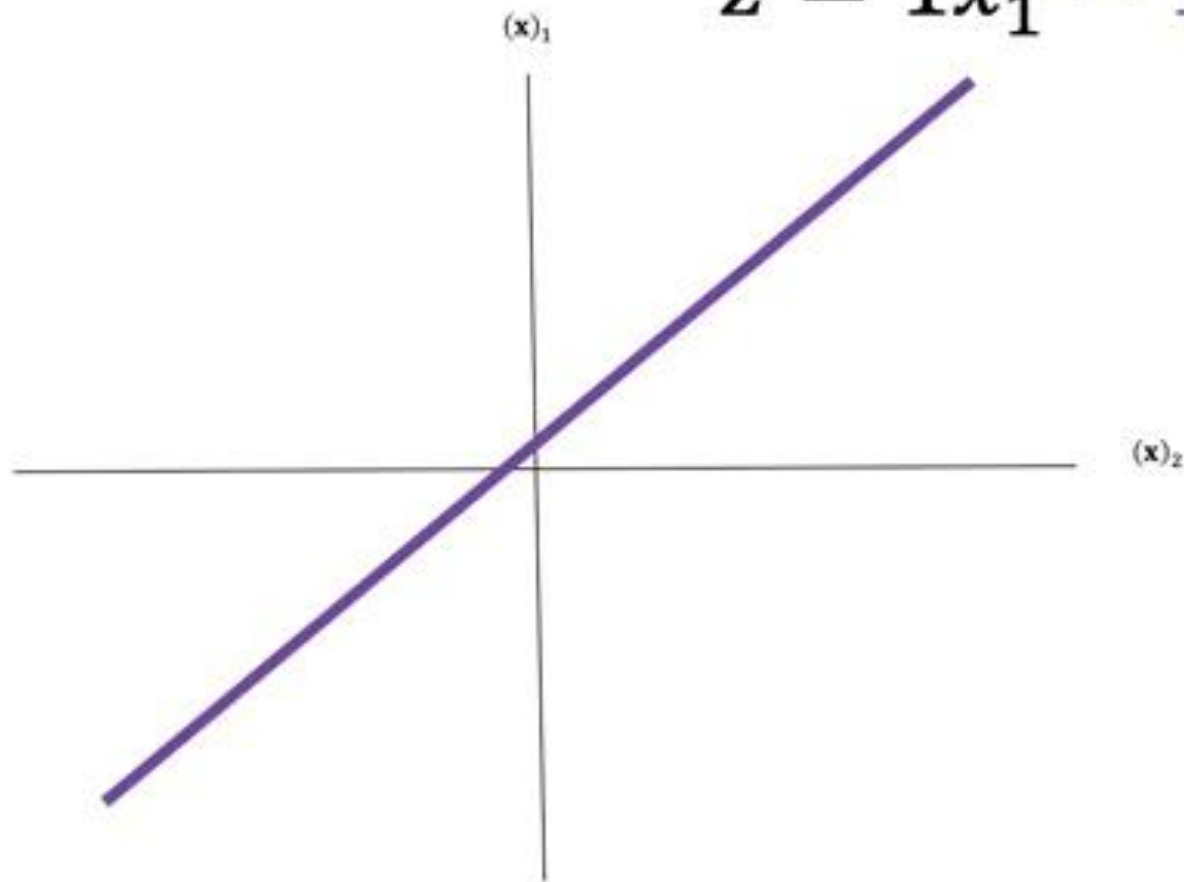


$$z = w_1x_1 + w_2x_2 + b$$

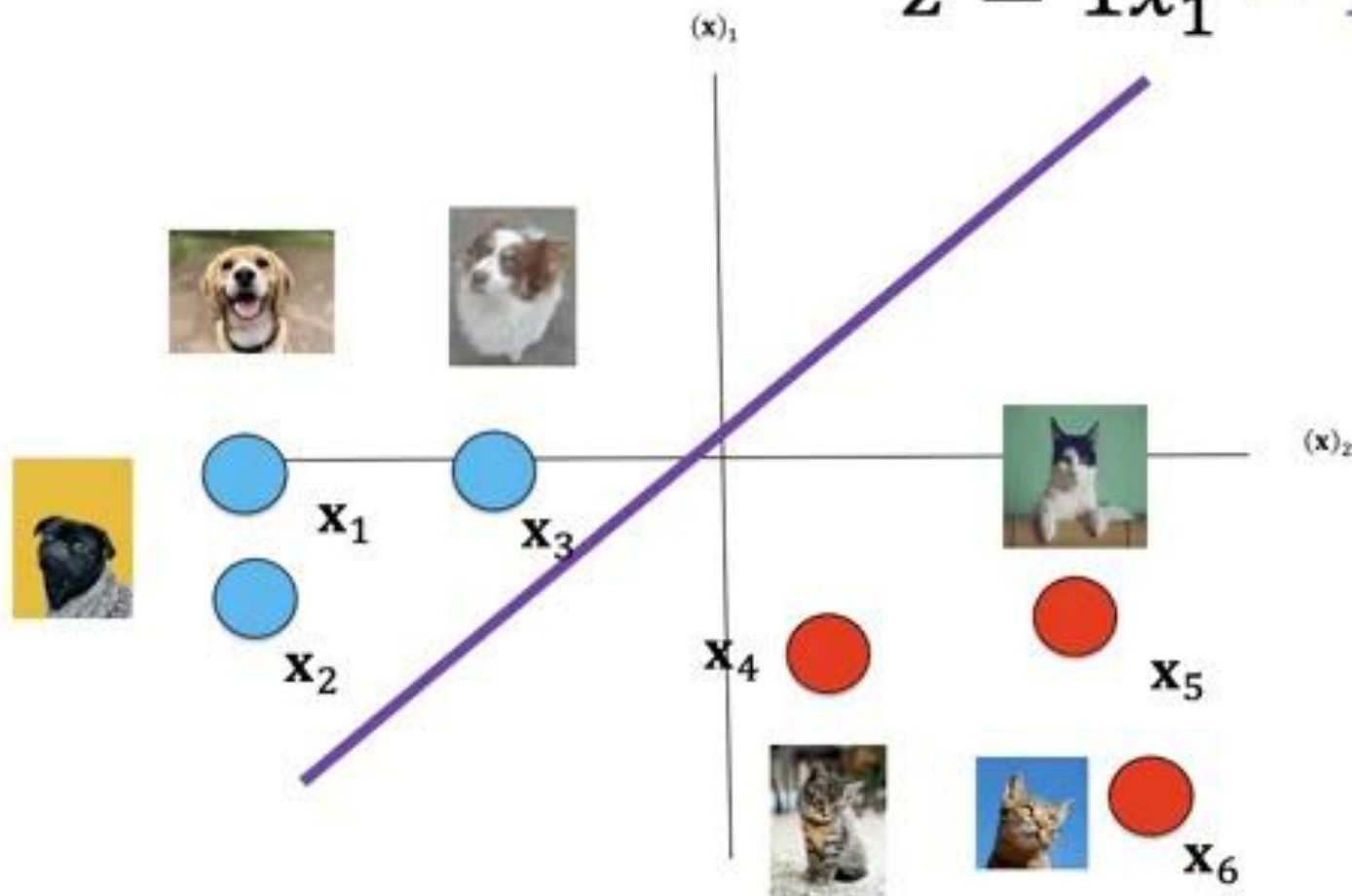
$$z = \mathbf{w}\mathbf{x} + b$$

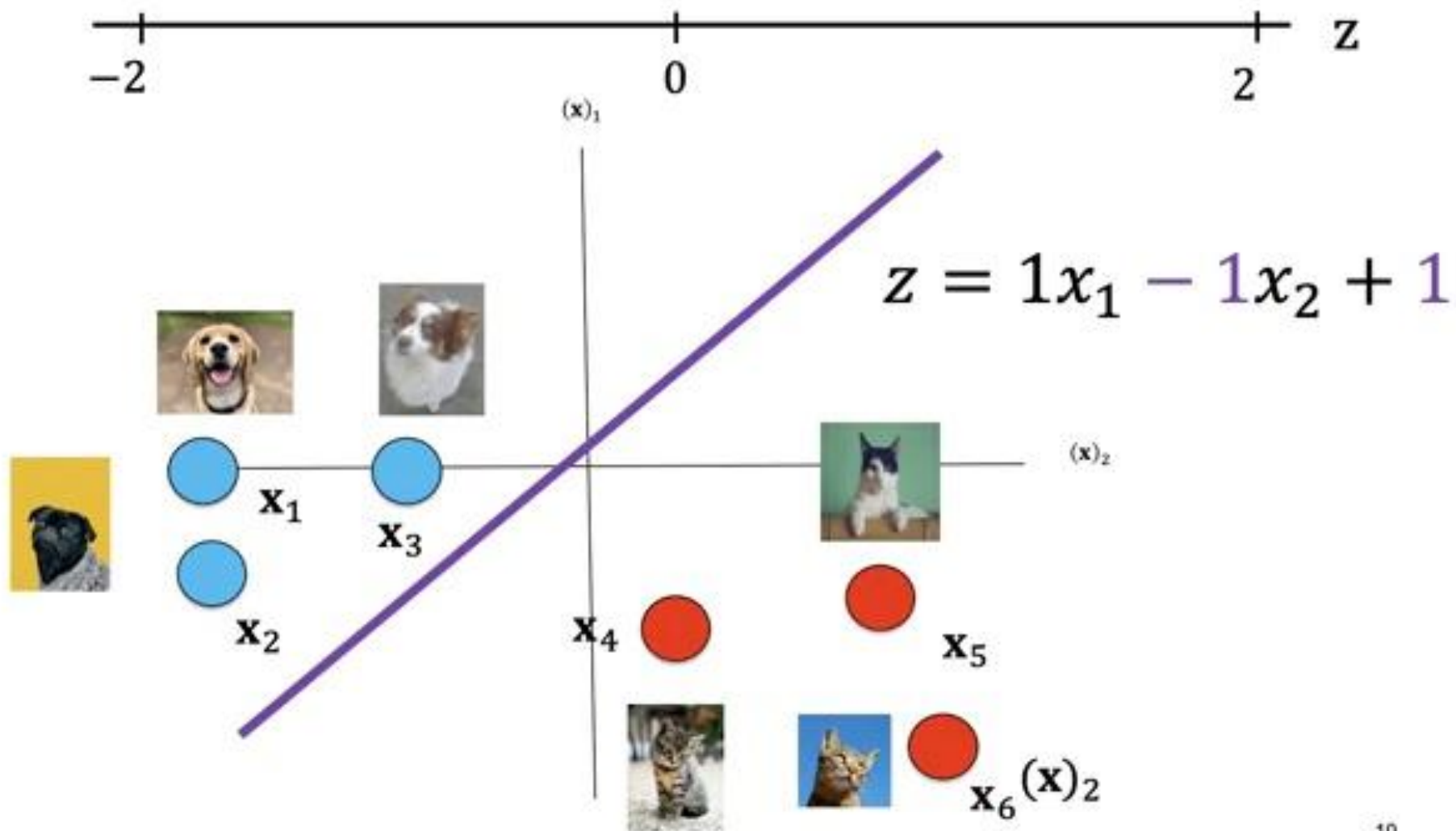


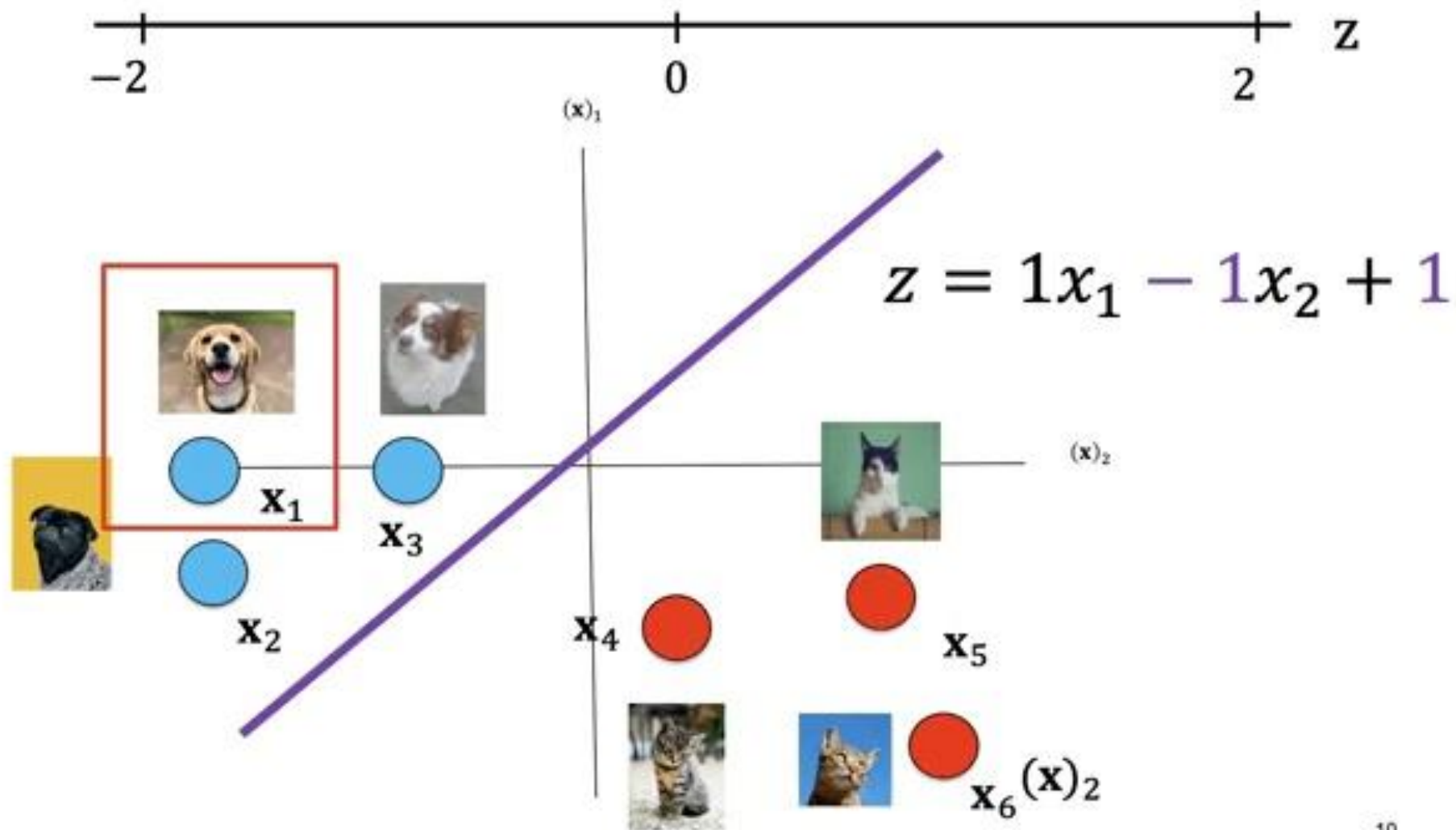
$$z = 1x_1 - 1x_2 + 1 = 0$$

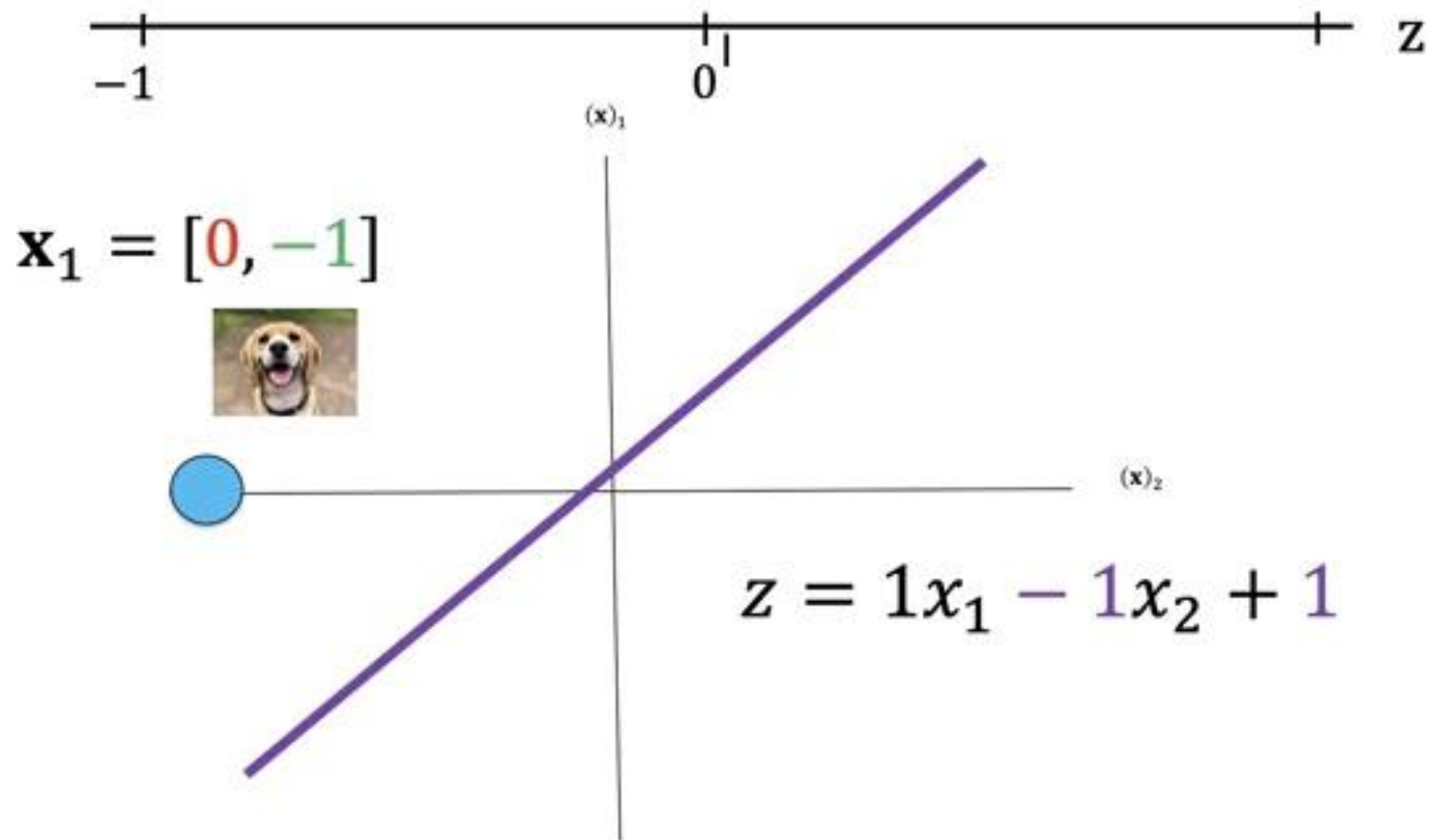


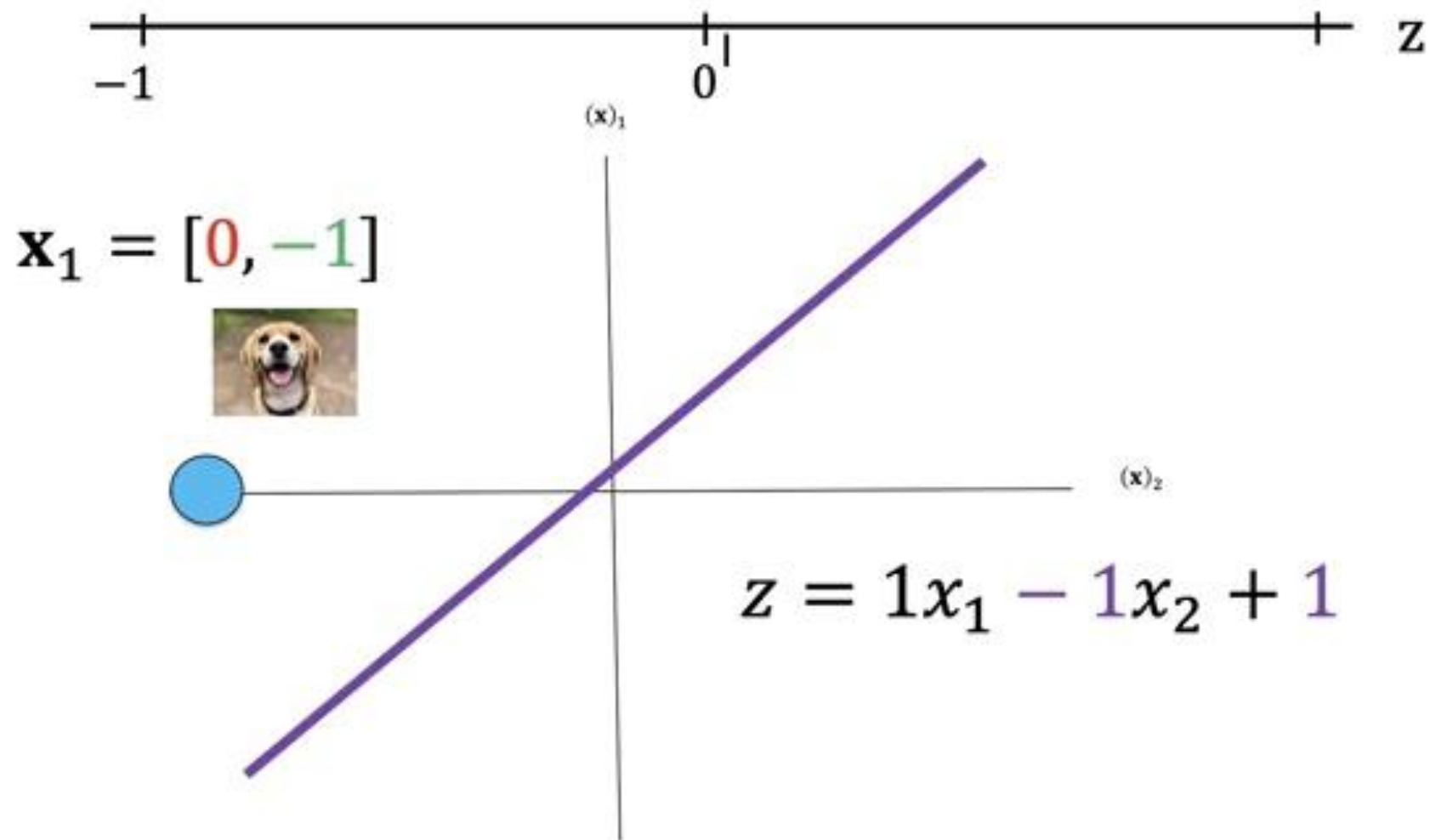
$$z = 1x_1 - 1x_2 + 1 = 0$$

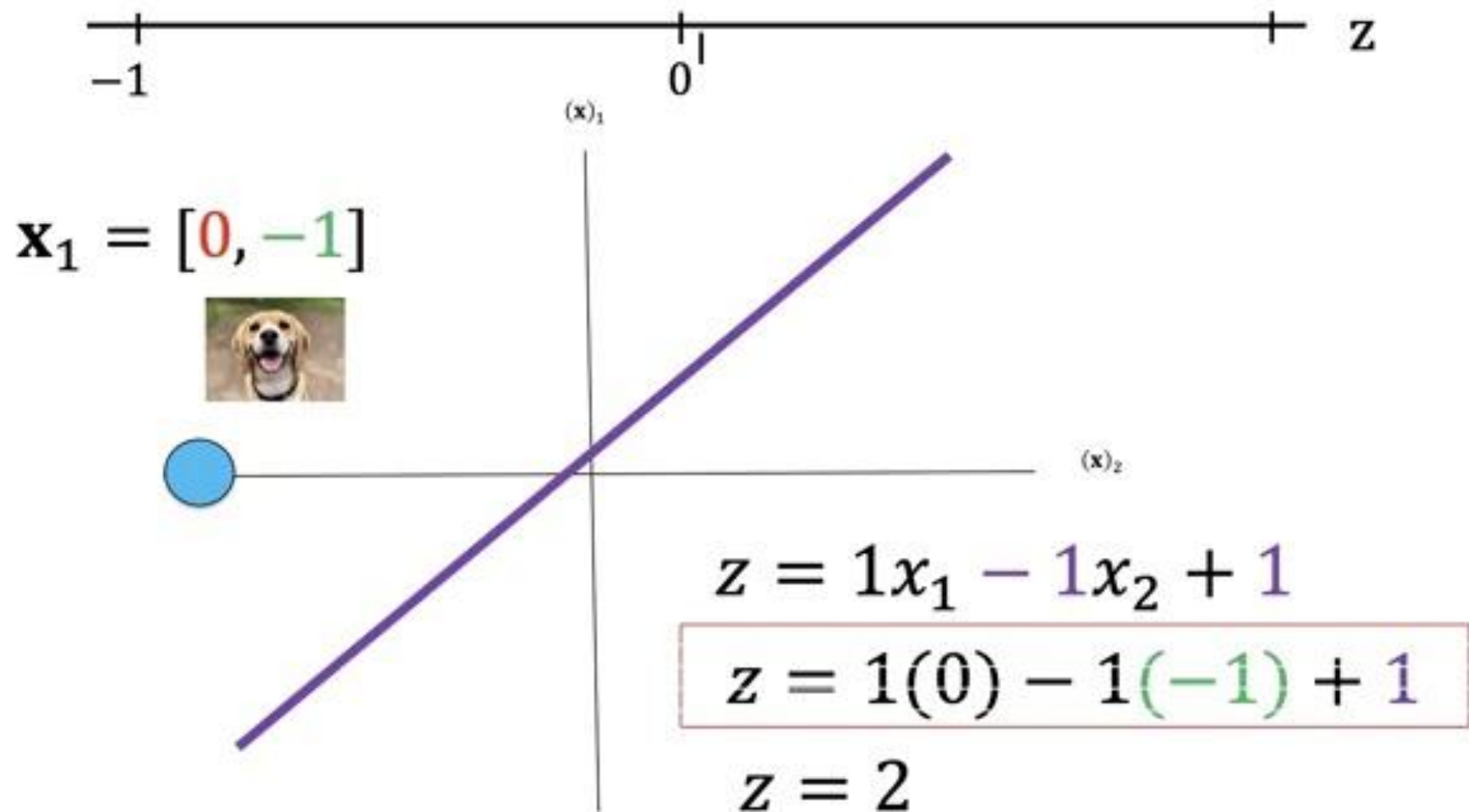


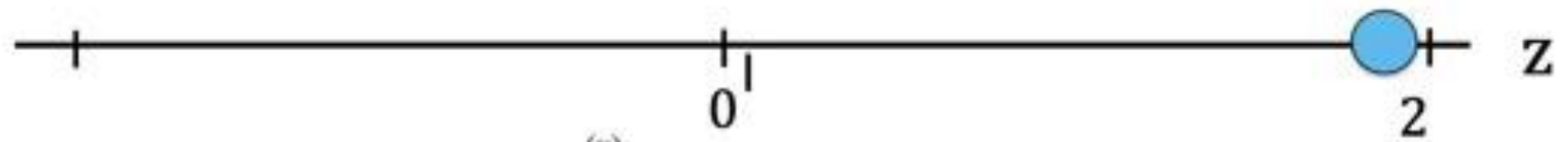




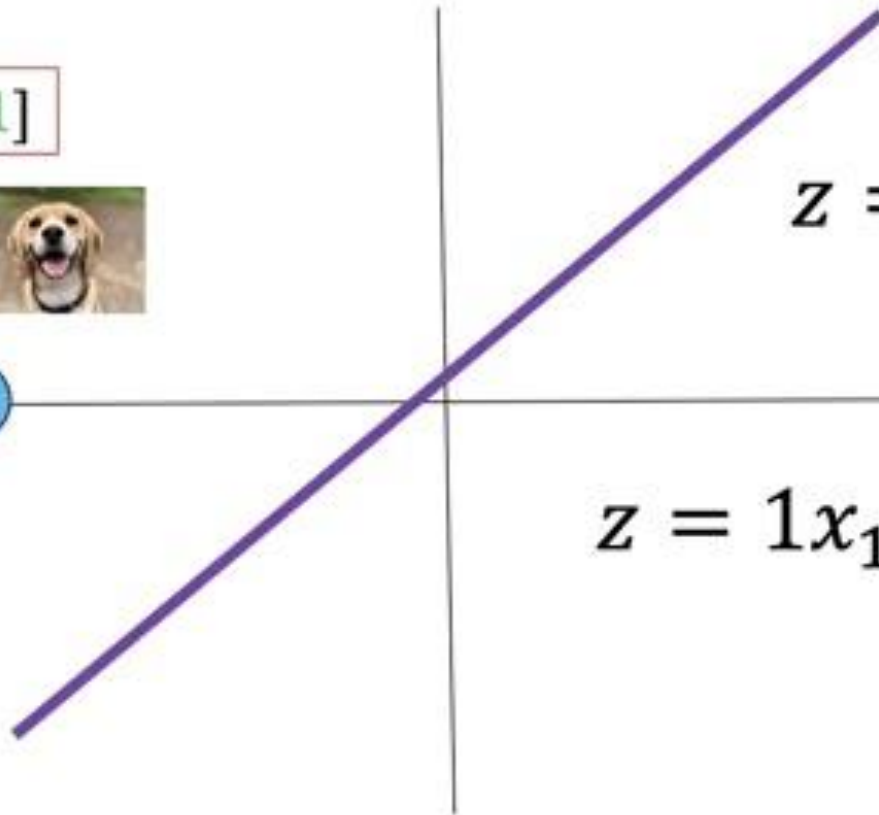








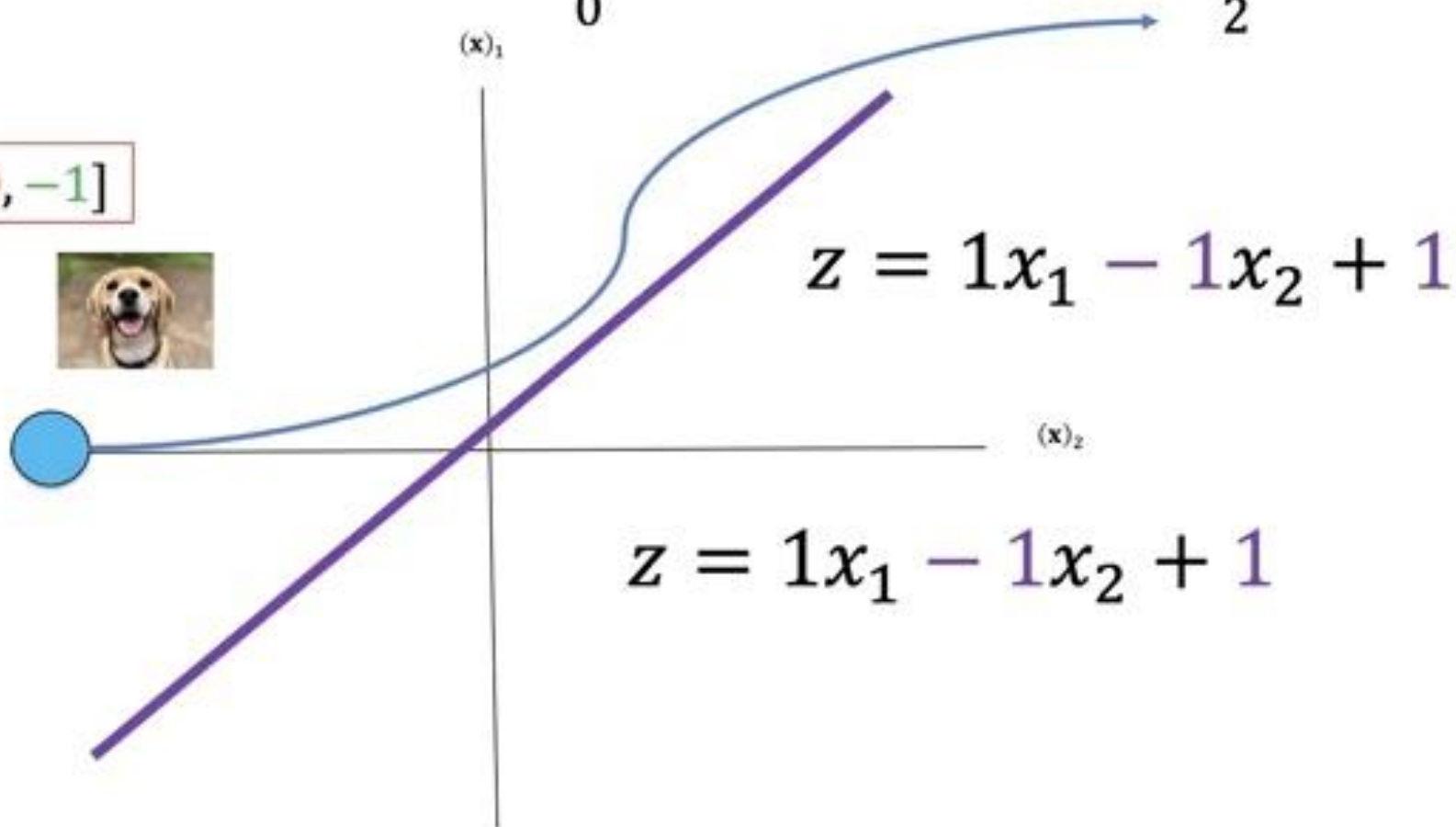
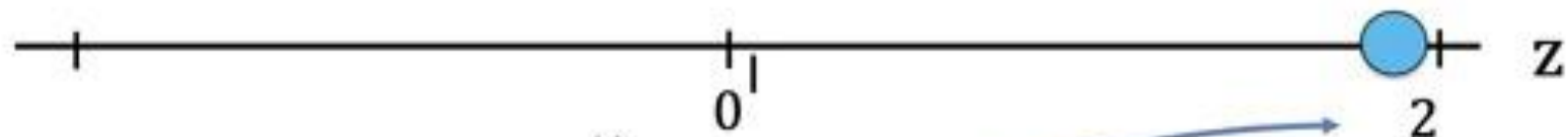
$$\mathbf{x}_1 = [0, -1]$$

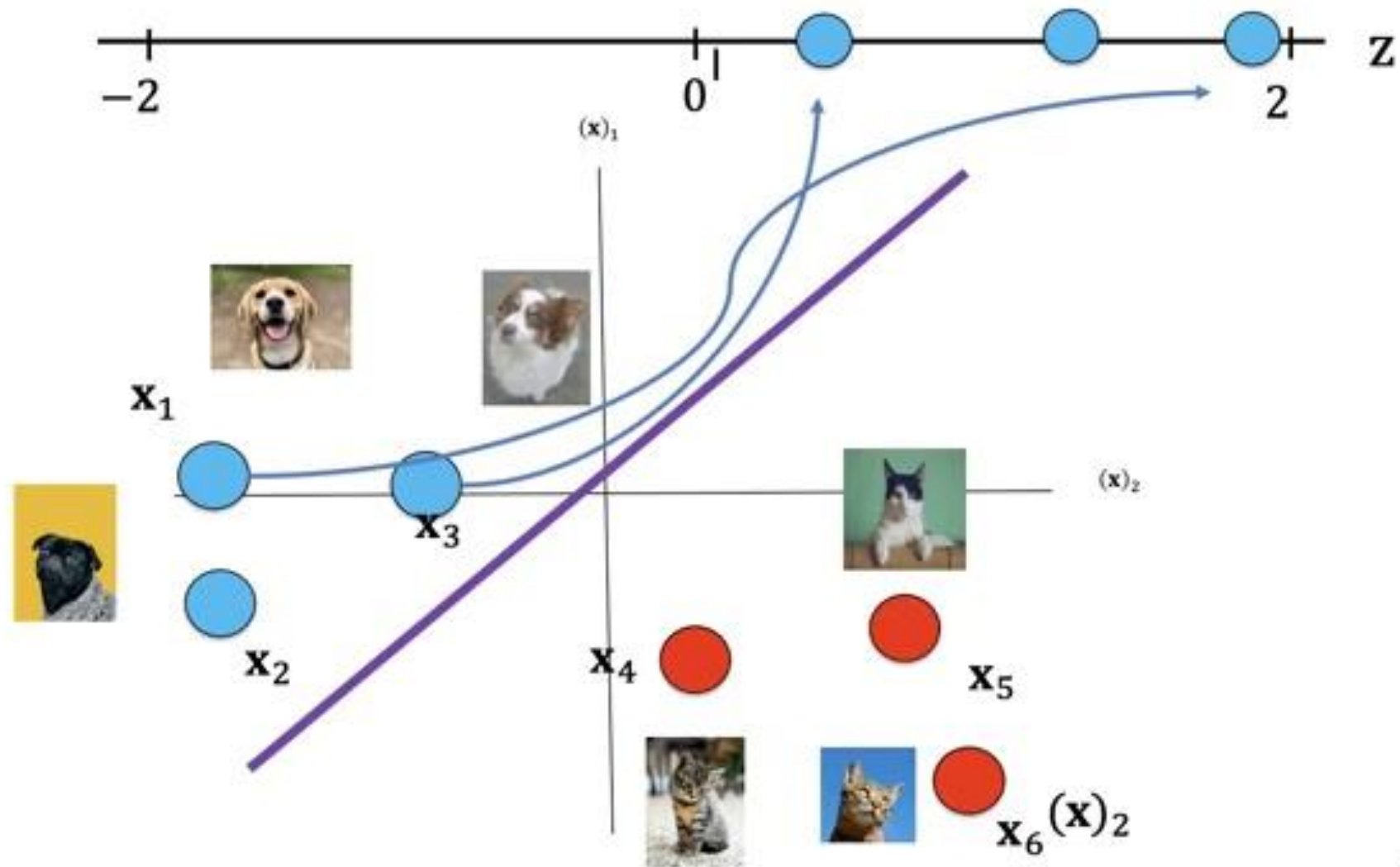


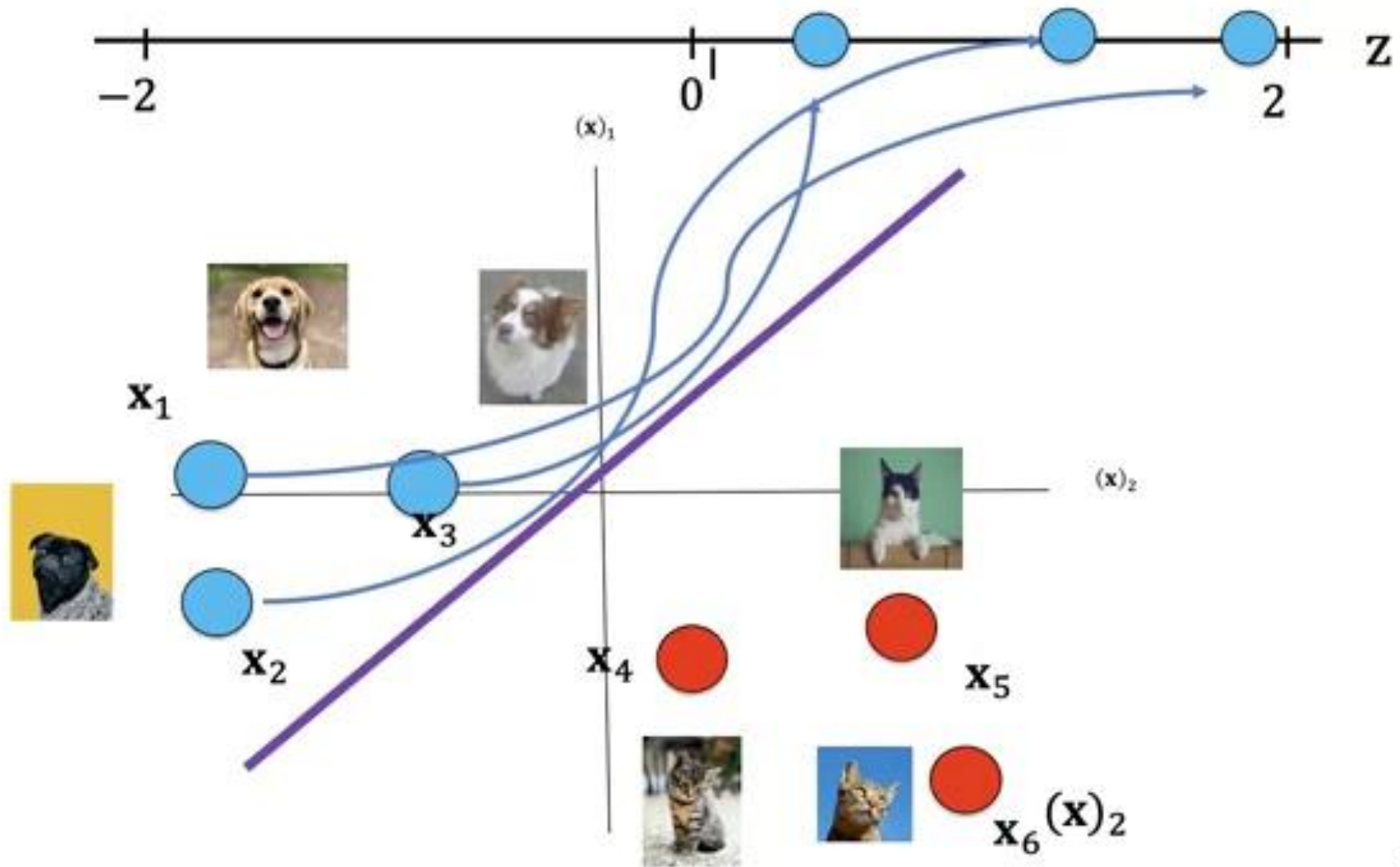
$$z = 1x_1 - 1x_2 + 1$$

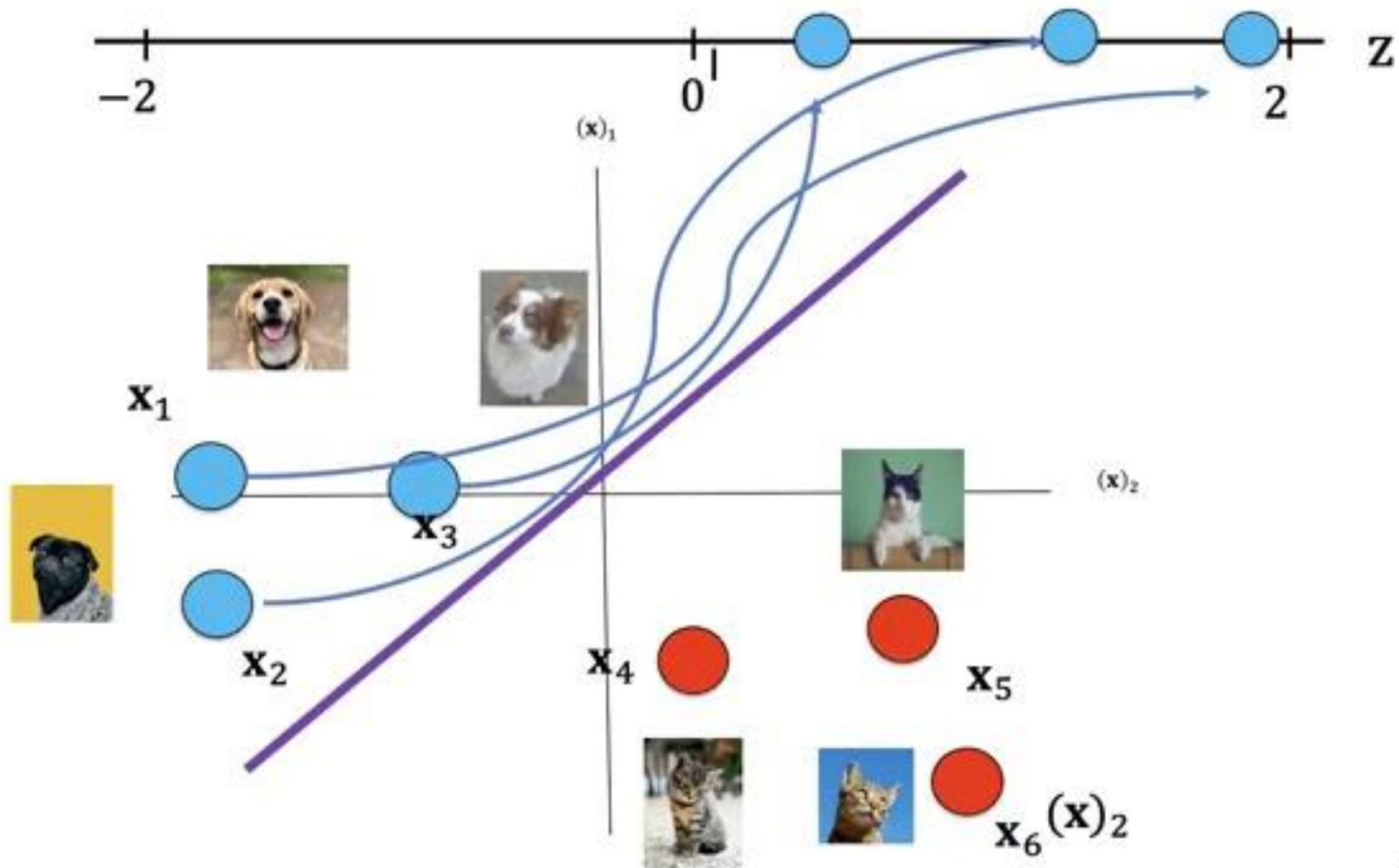
$$z = 1x_1 - 1x_2 + 1$$

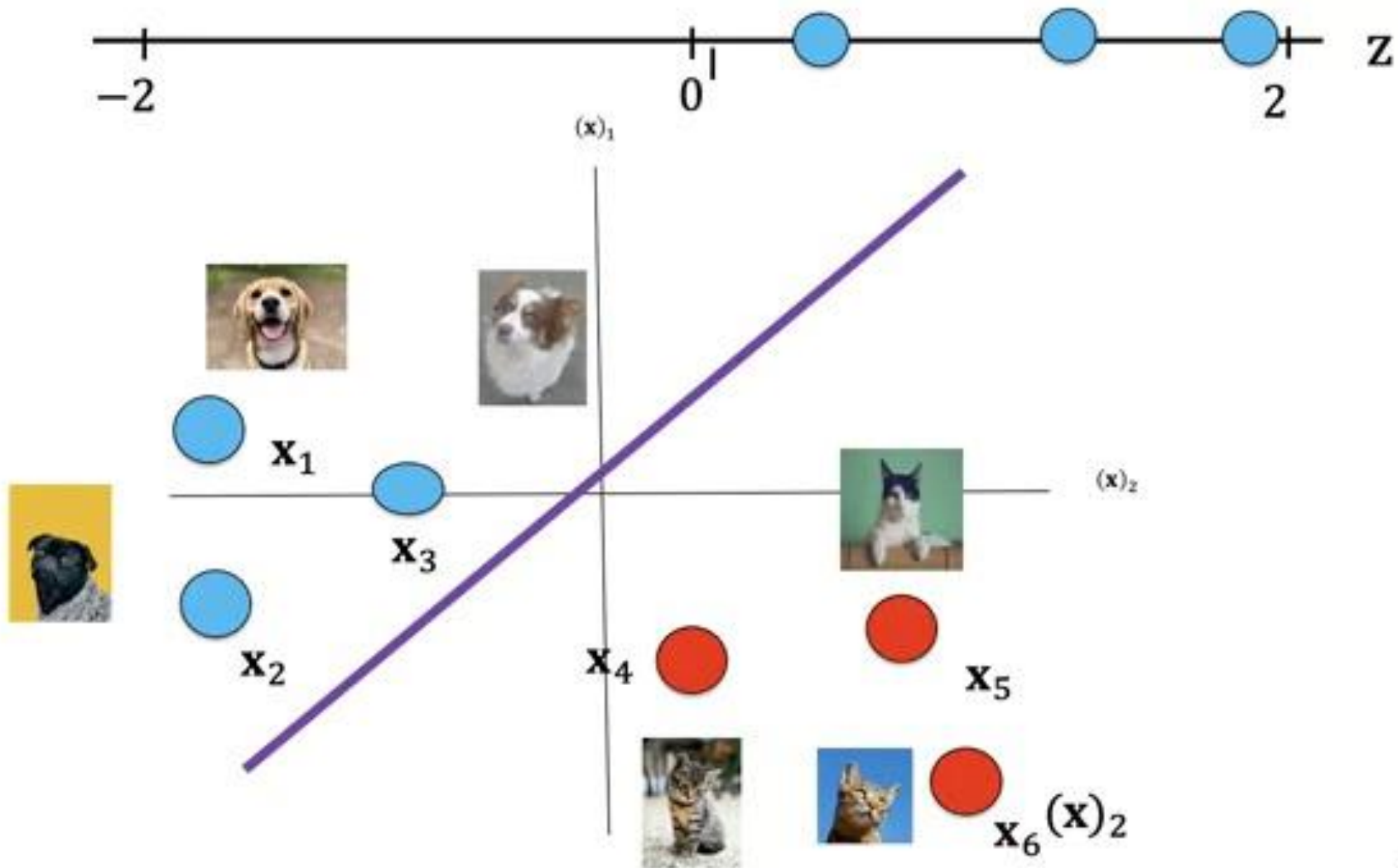
$$\mathbf{x}_1 = [0, -1]$$

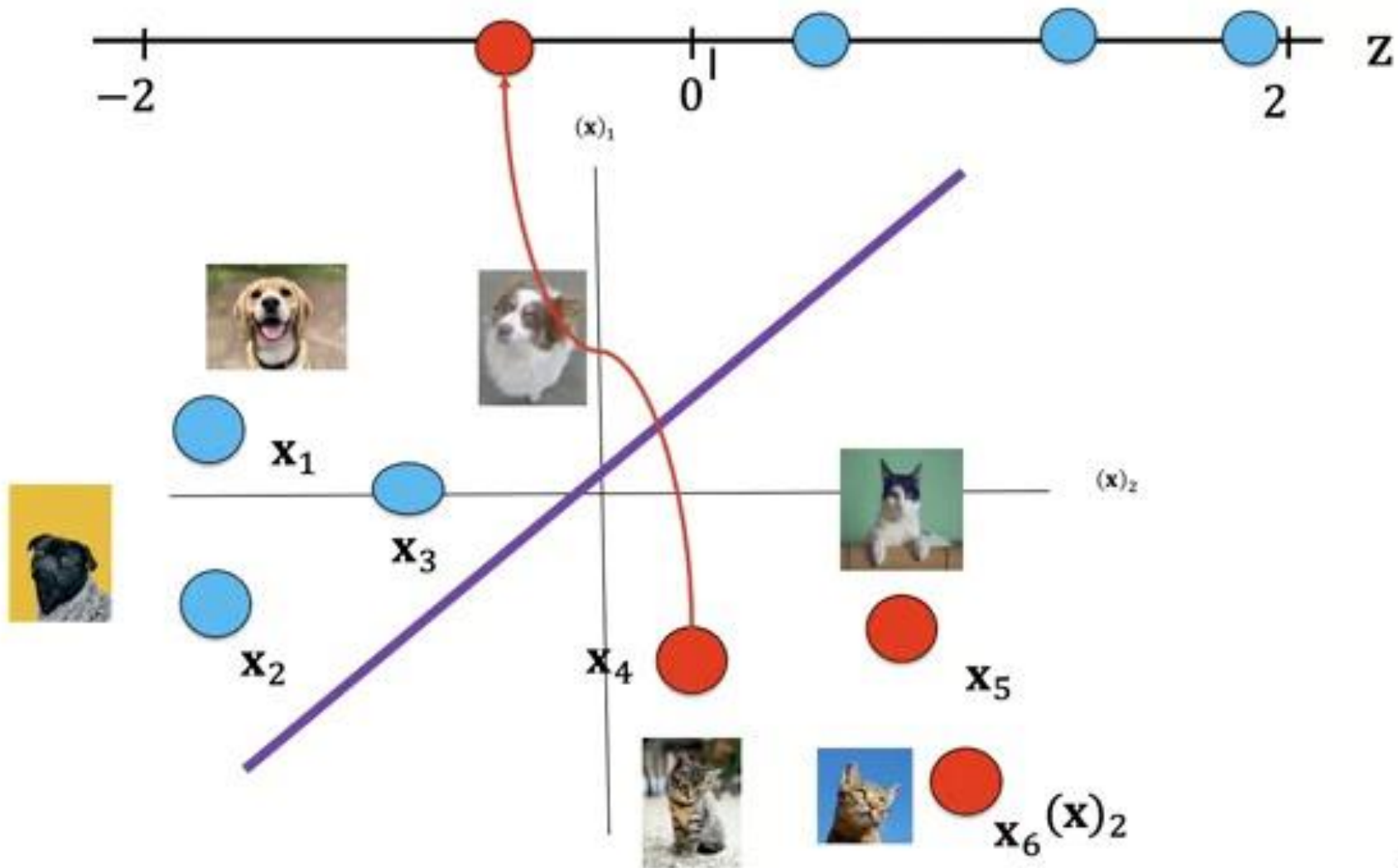


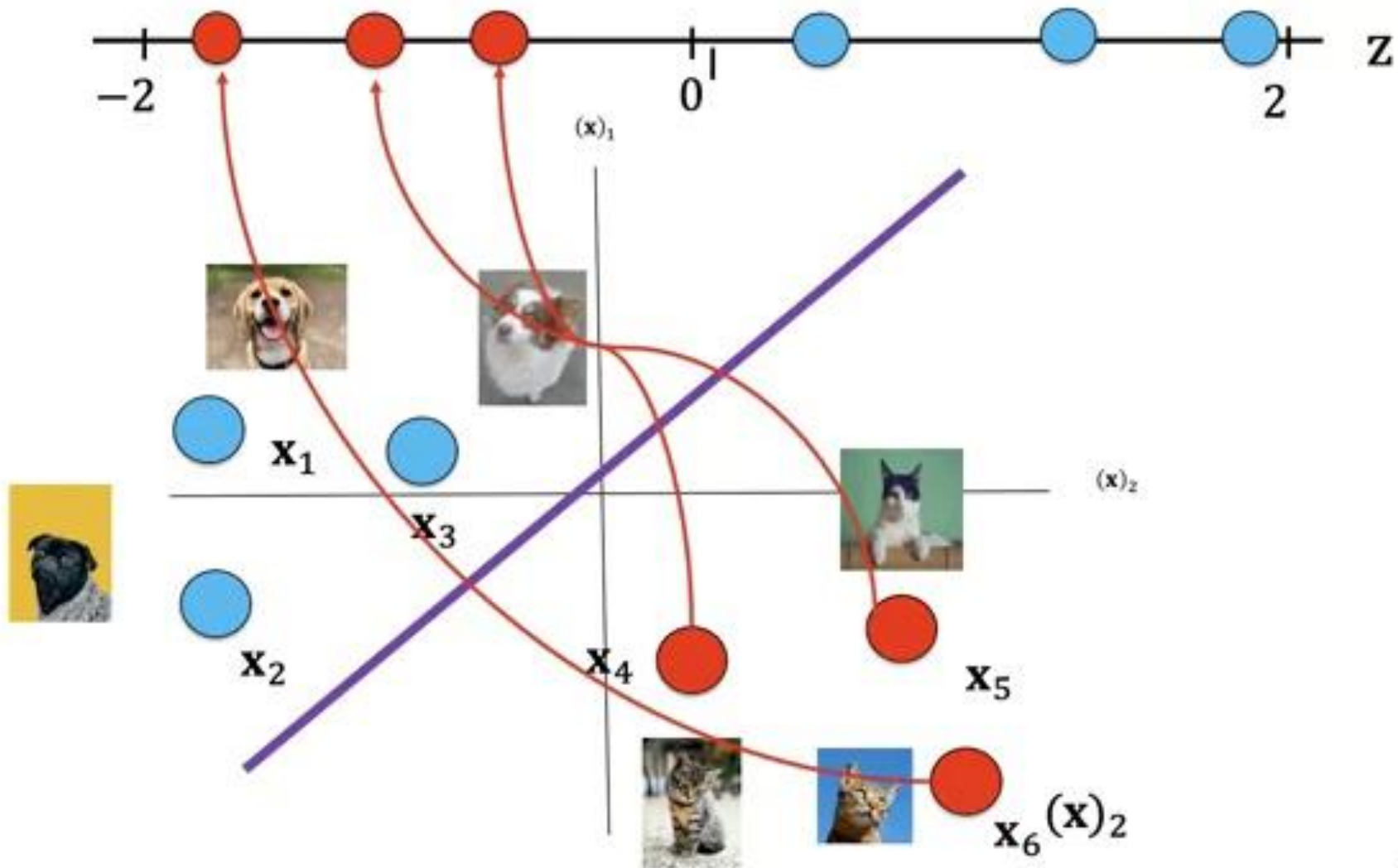




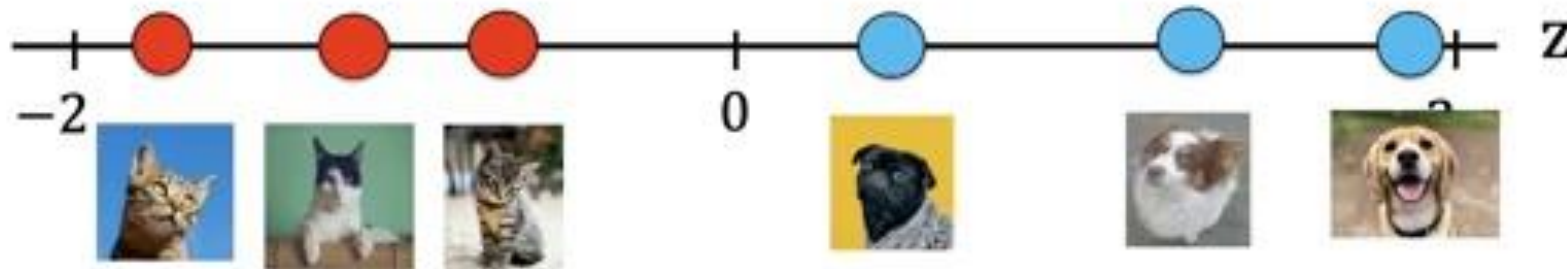




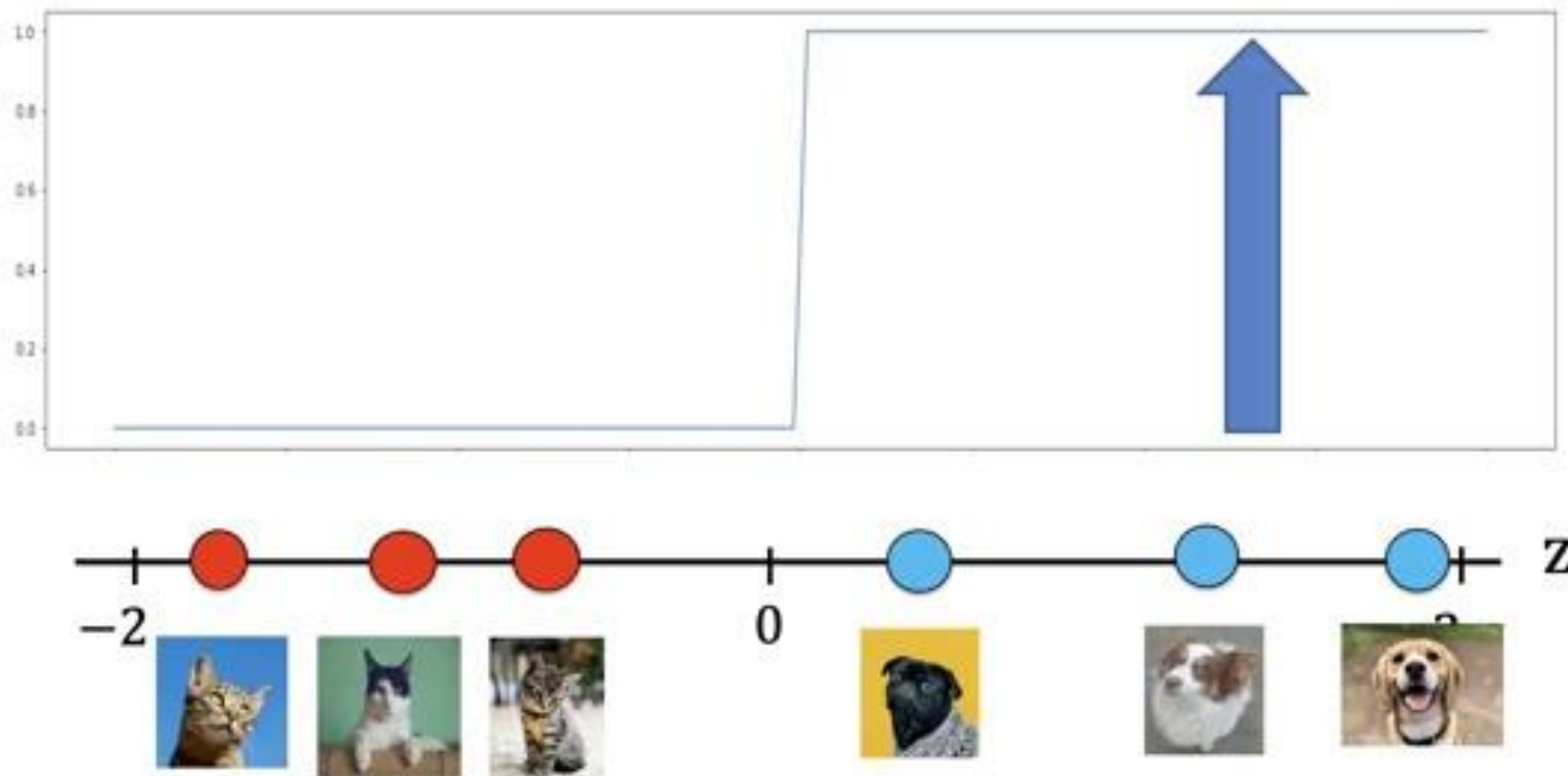




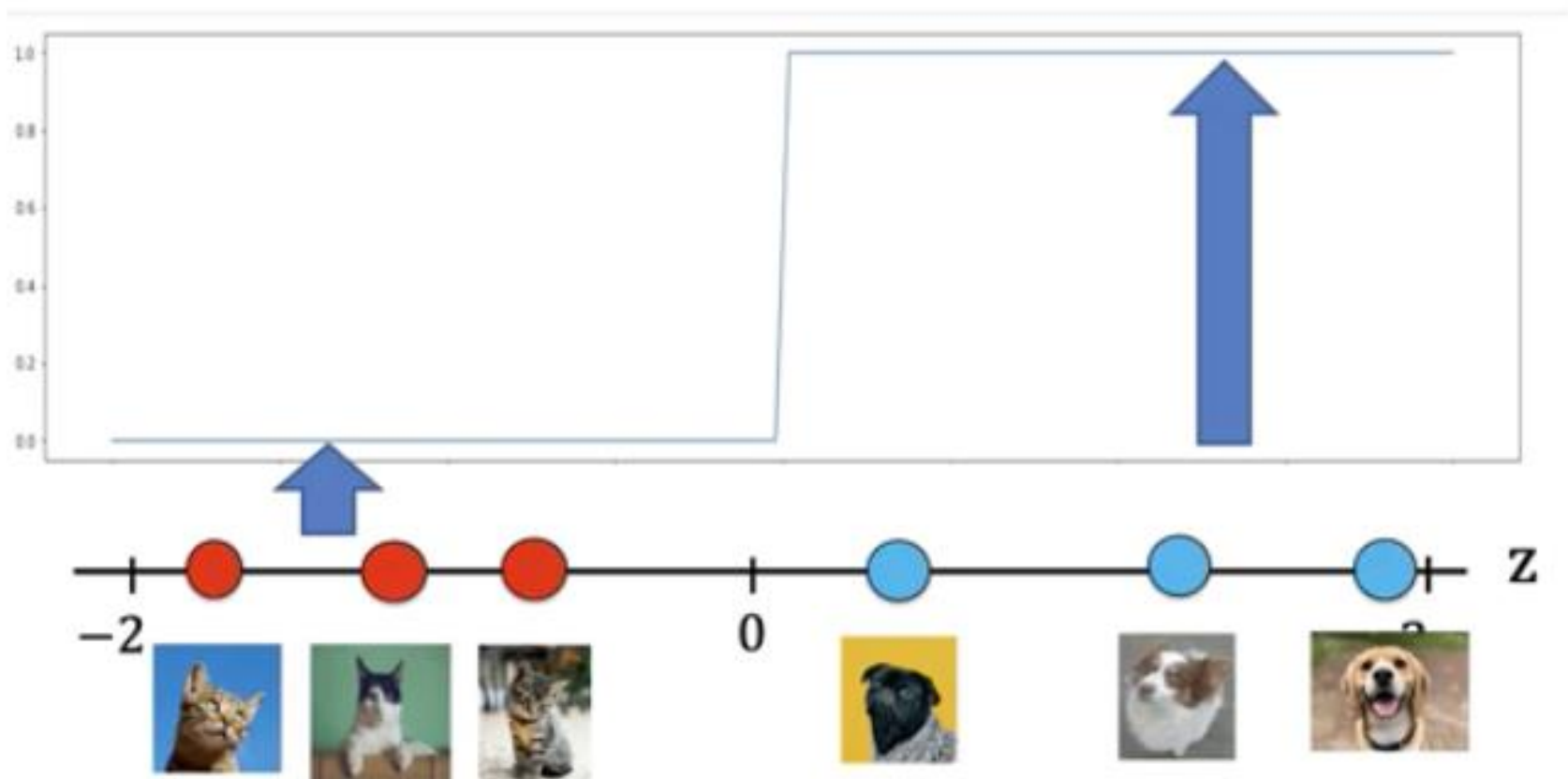
$$\hat{y} = f(z)$$



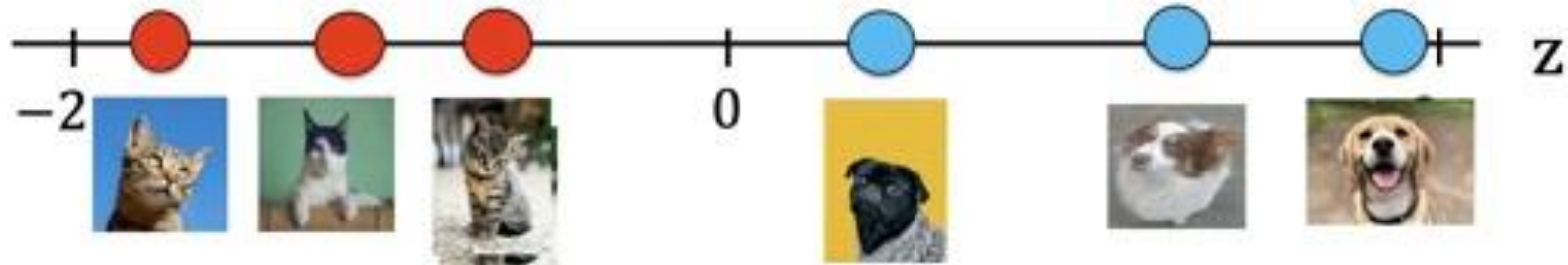
$$\hat{y} = f(z)$$



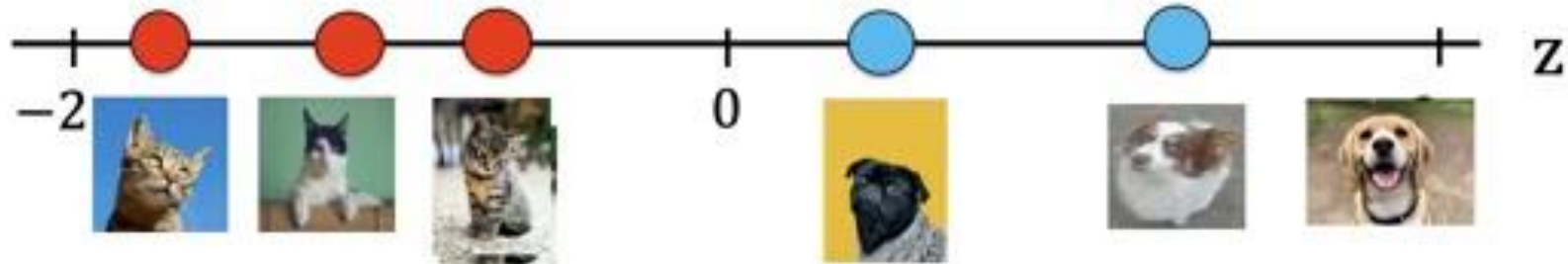
$$\hat{y} = f(z)$$



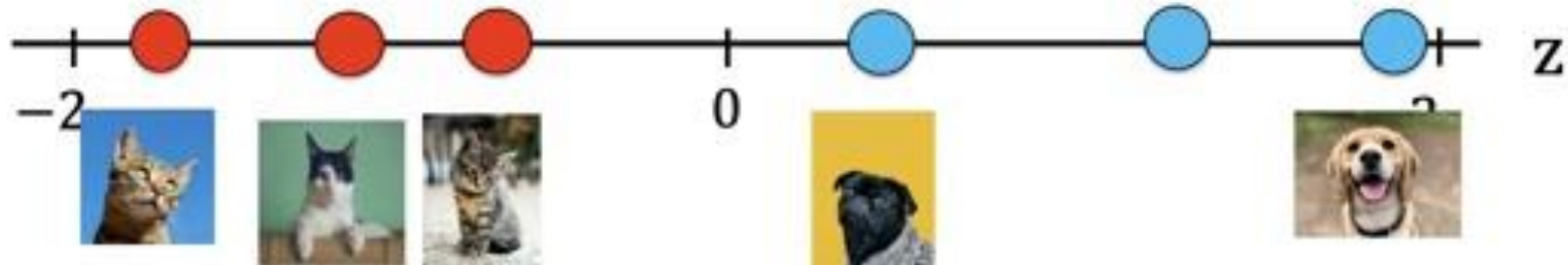
$$\hat{y}_1 = f(\text{dog image})$$



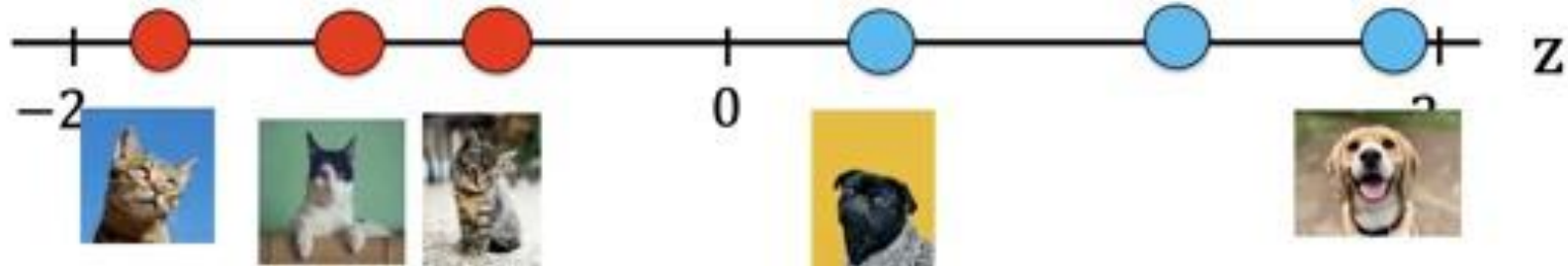
$$\hat{y}_1 = f(\text{dog image}) = 1$$



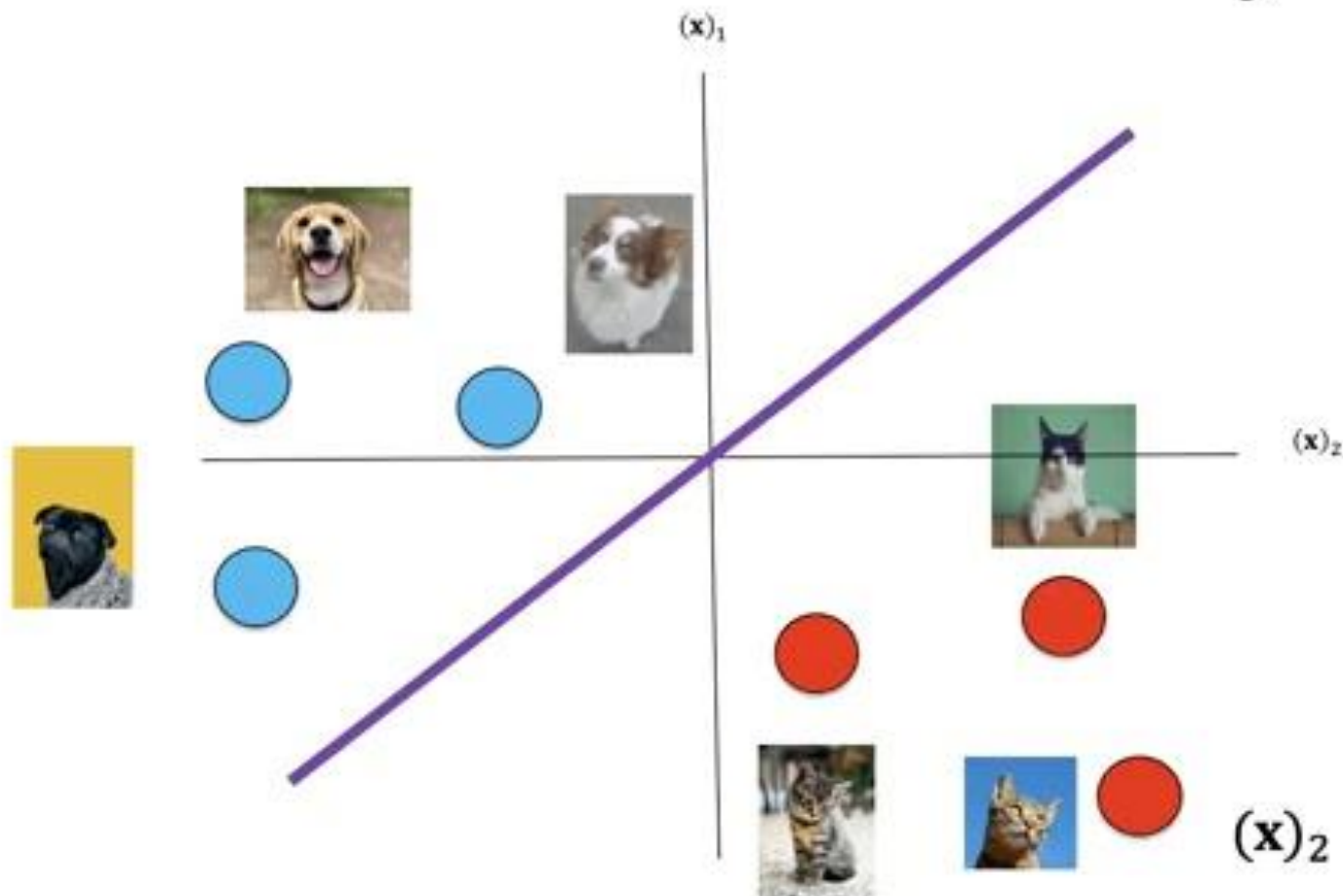
$$\hat{y}_1 = f(\text{cat image})$$



$$\hat{y}_1 = f(\text{cat image}) = 0$$

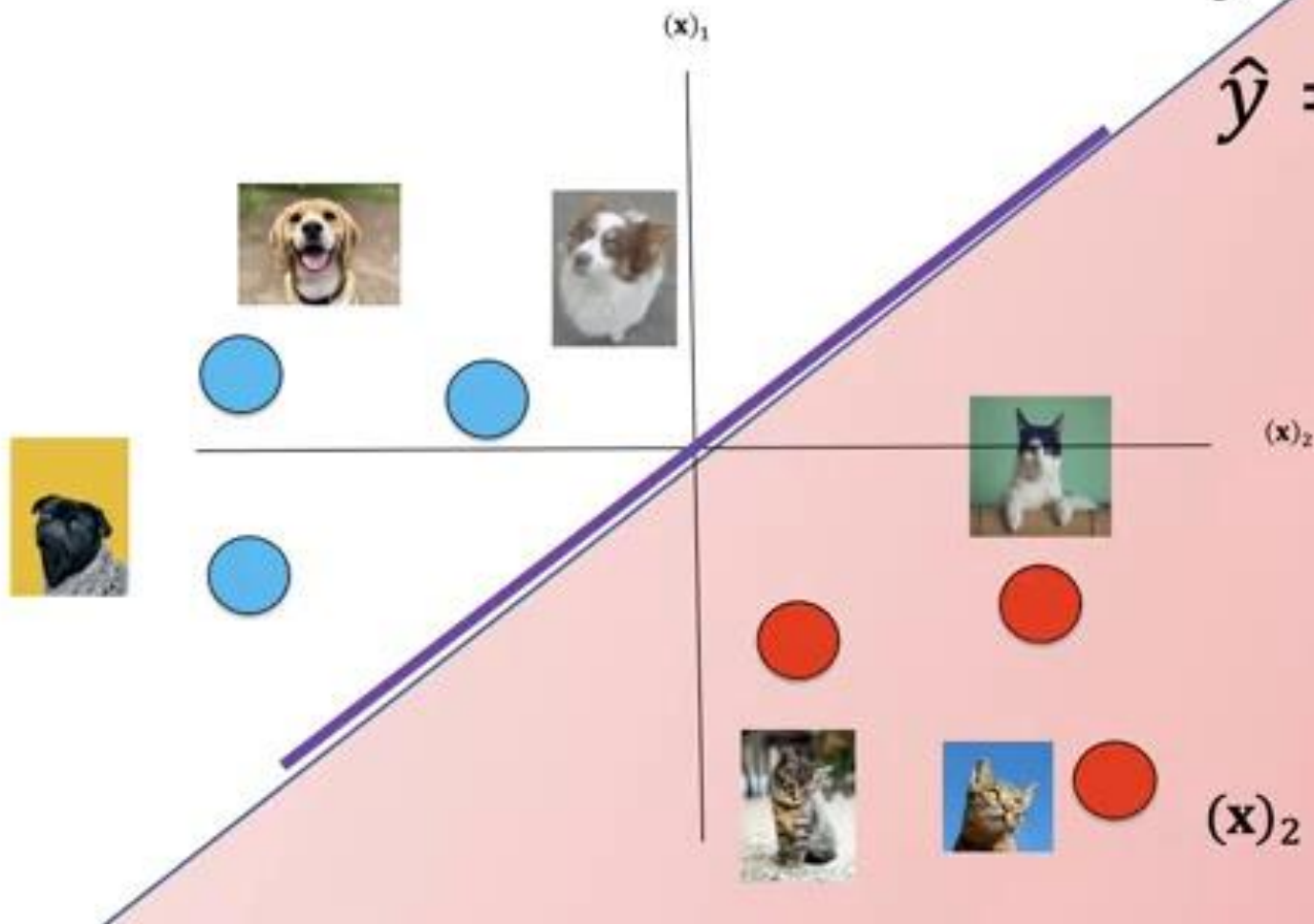


$$\hat{y} = f(\mathbf{w}\mathbf{x} + \mathbf{b})$$

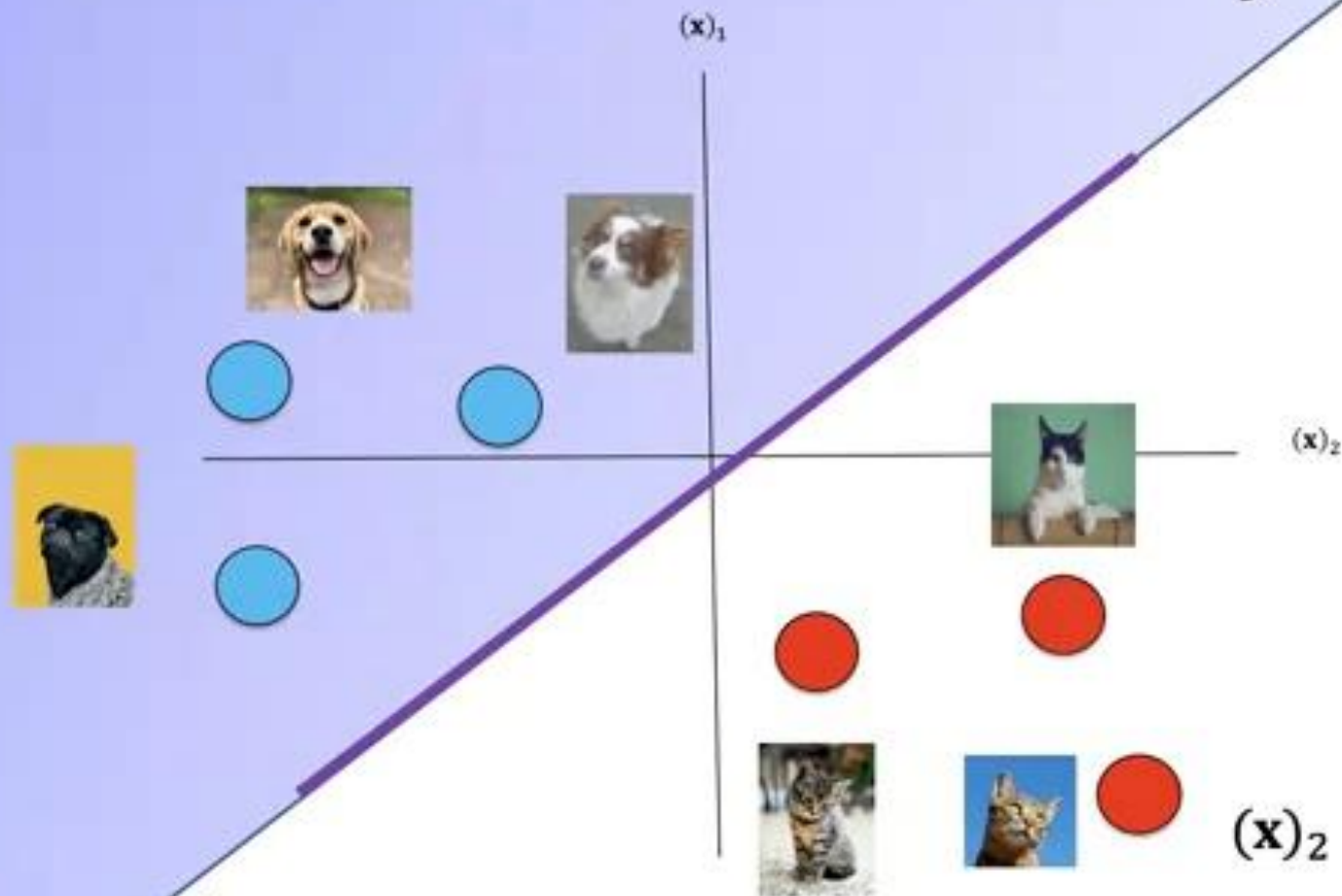


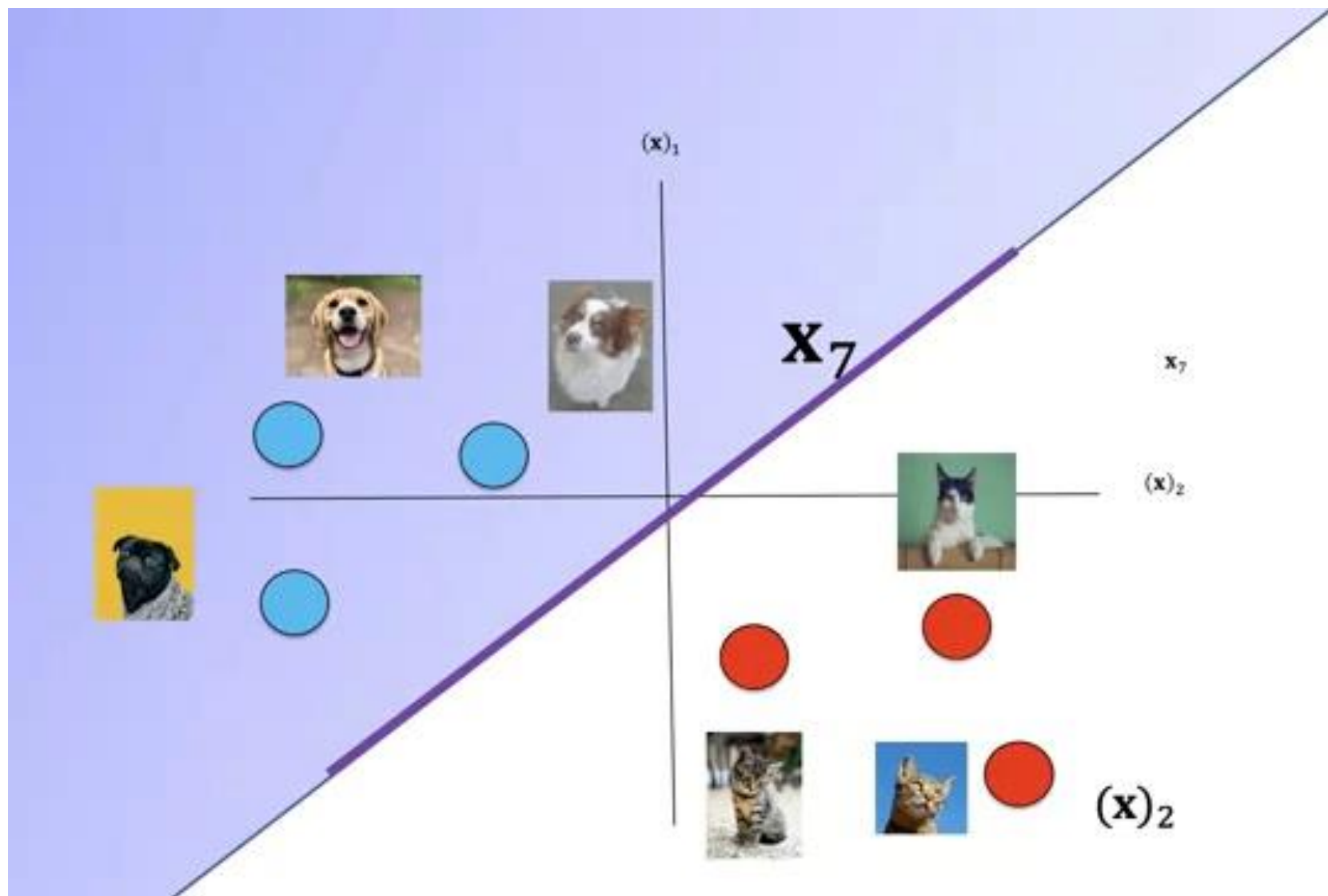
$$\hat{y} = f(\mathbf{w}\mathbf{x} + \mathbf{b})$$

$$\hat{y} = 0$$



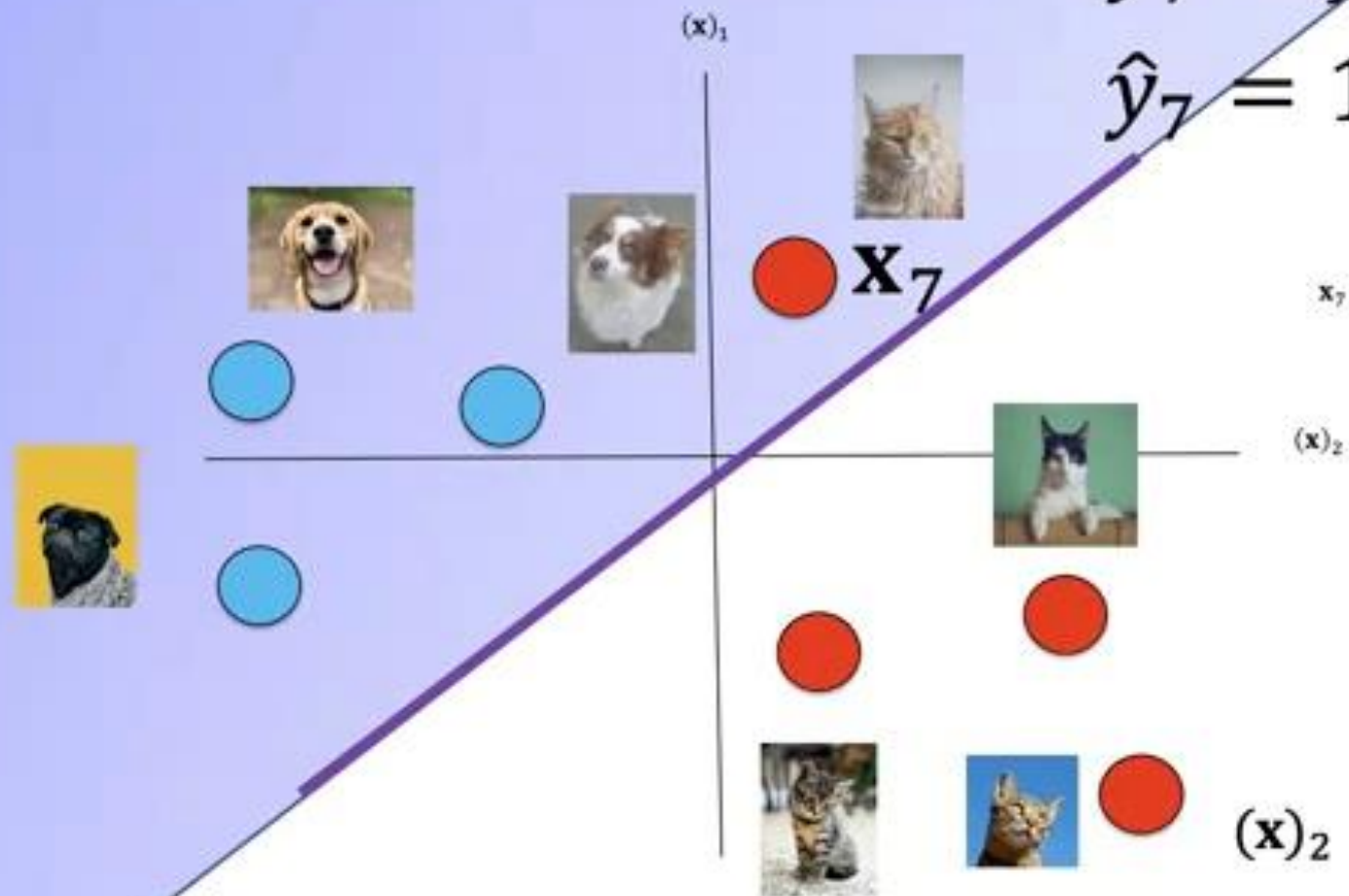
$$\hat{y} = f(\mathbf{w}\mathbf{x} + \mathbf{b})$$





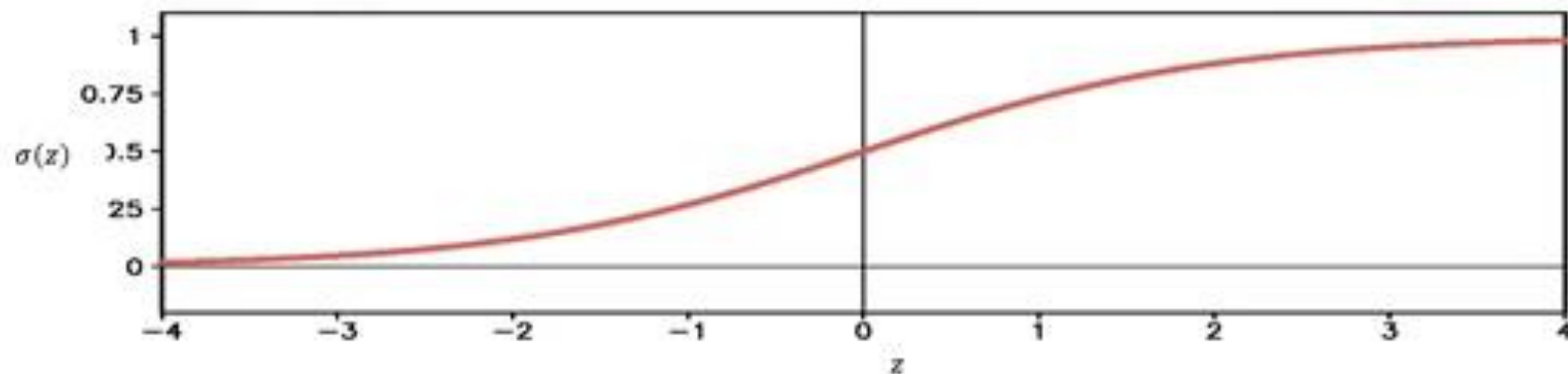
$$\hat{y}_7 = f(\mathbf{w}\mathbf{x}_7 + \mathbf{b})$$

$$\hat{y}_7 = 1$$



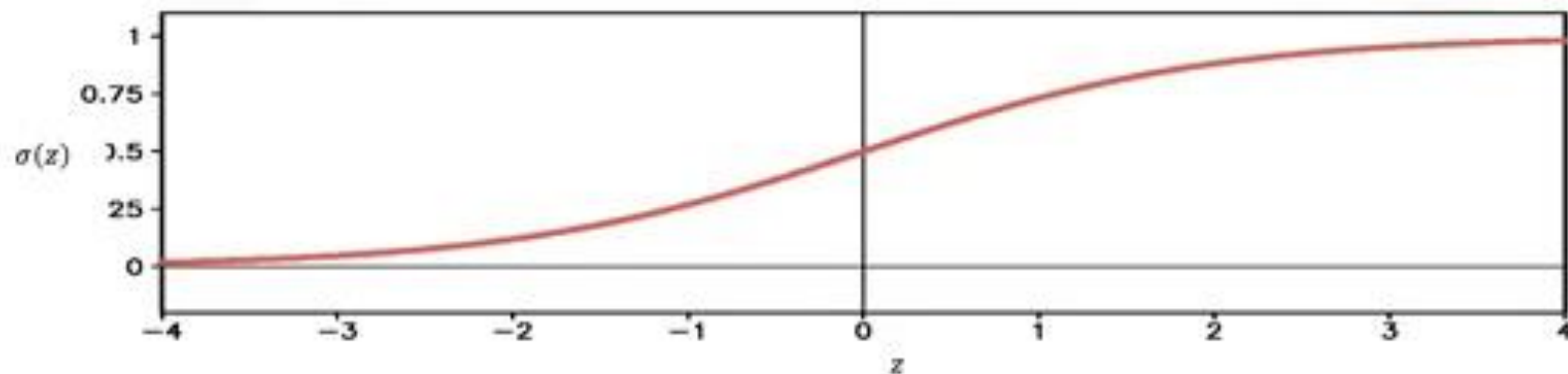
Logistic Regression

- Logistic Function



Logistic Regression

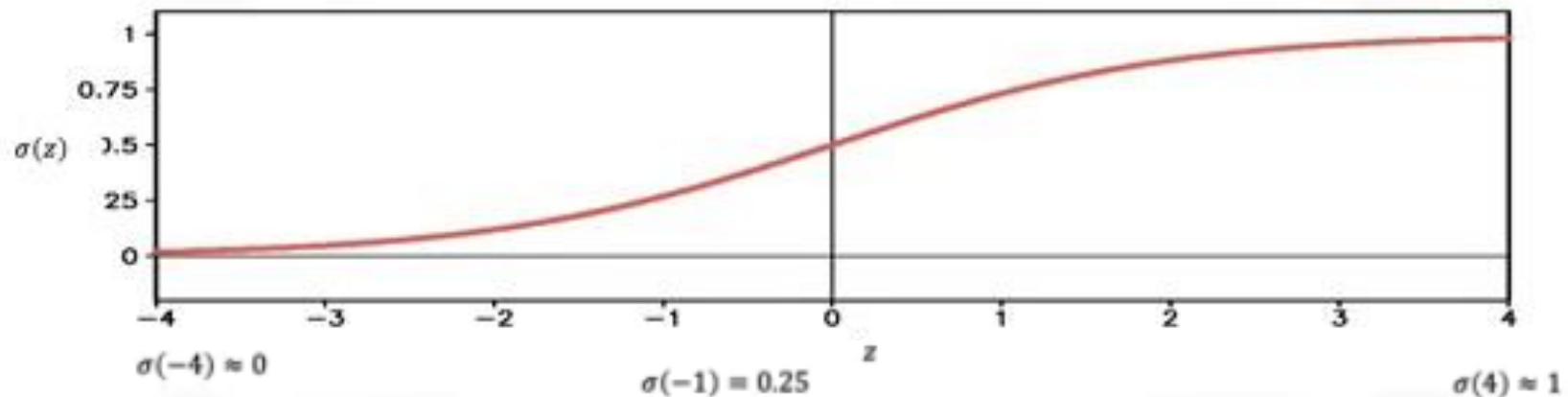
- Logistic Function



$$\sigma(z) = \frac{1}{1 + e^{-z}}$$

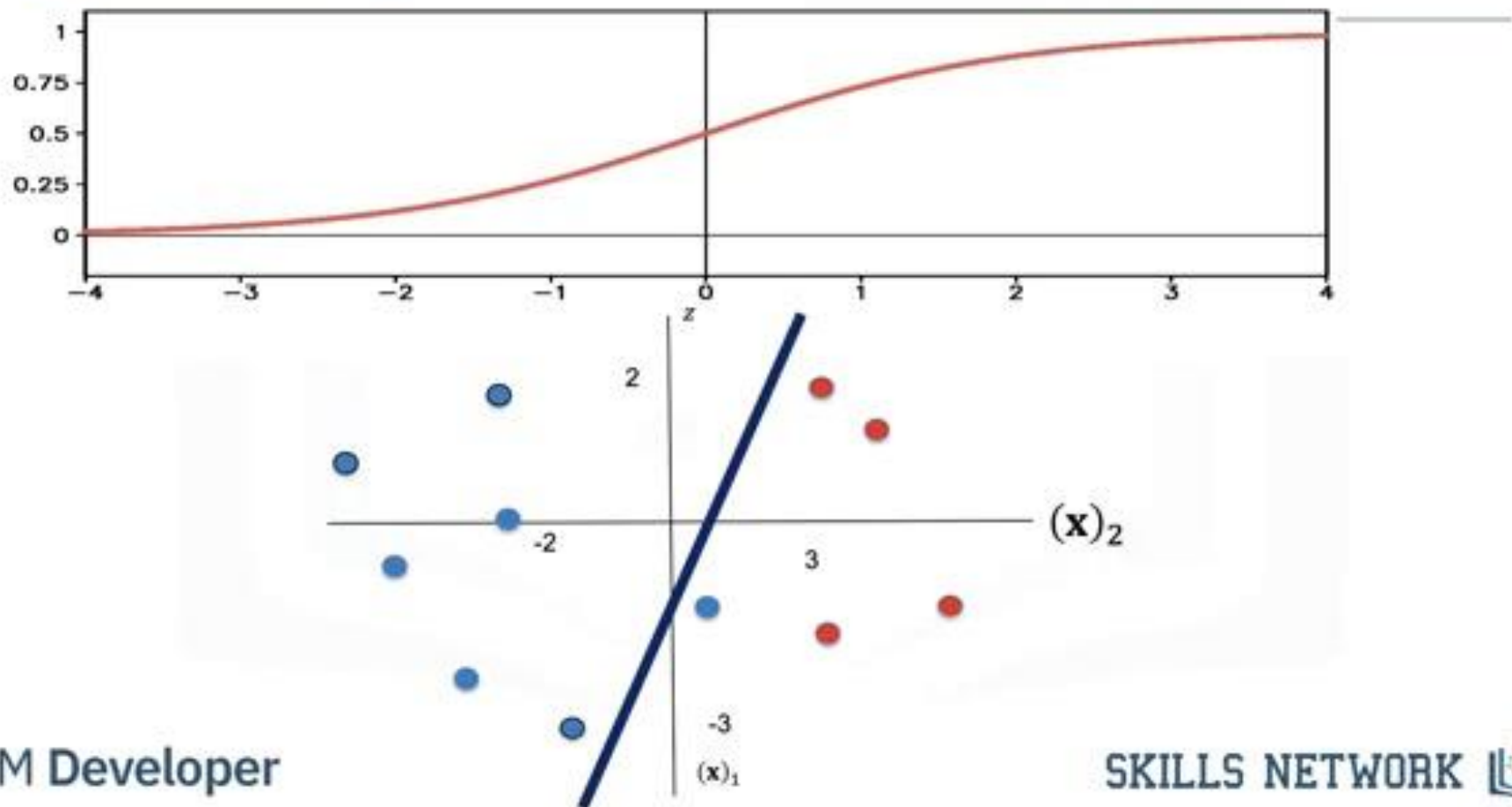
Logistic Regression

- Logistic Function

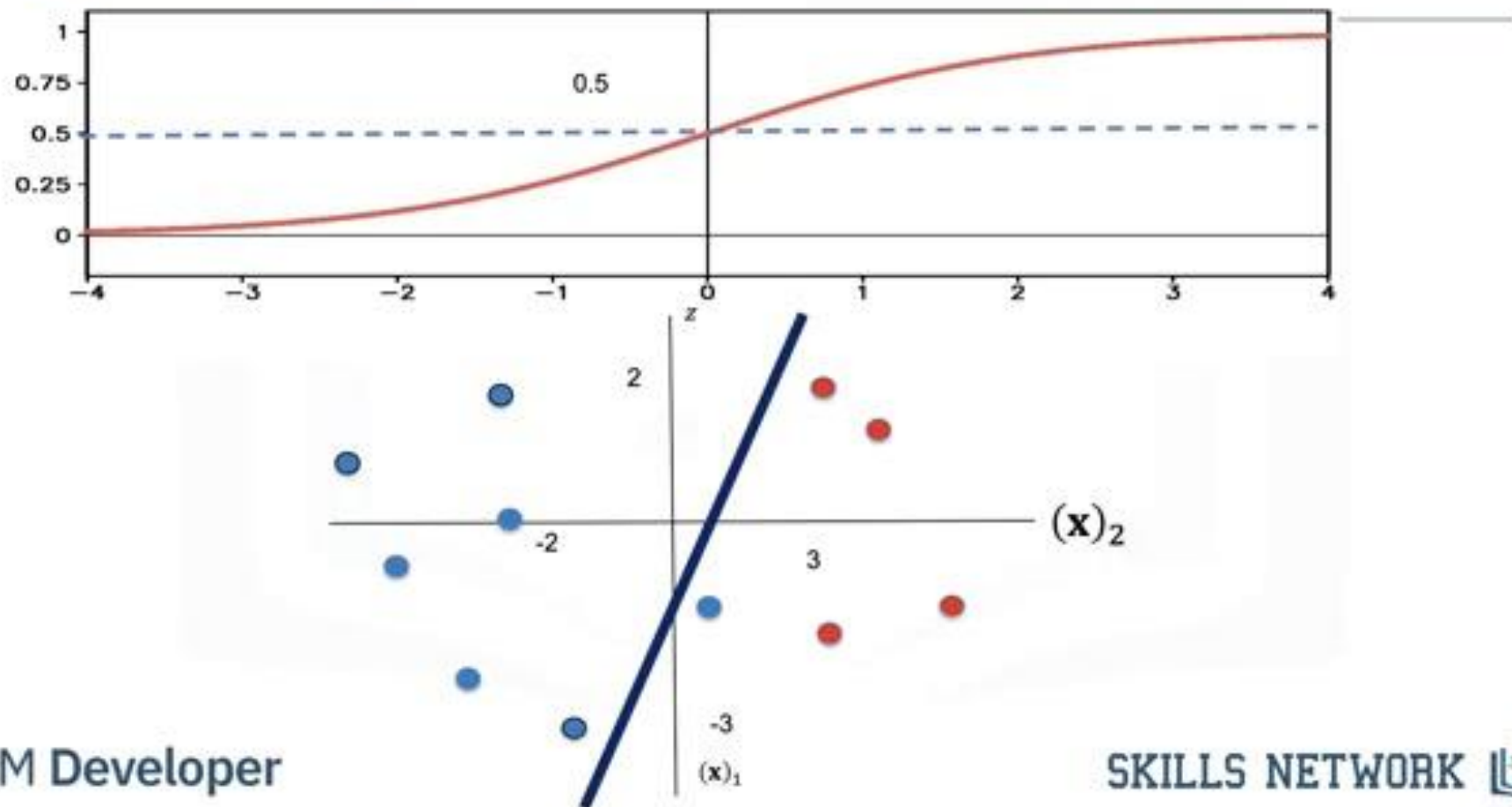


$$\sigma(z) = \frac{1}{1 + e^{-z}}$$

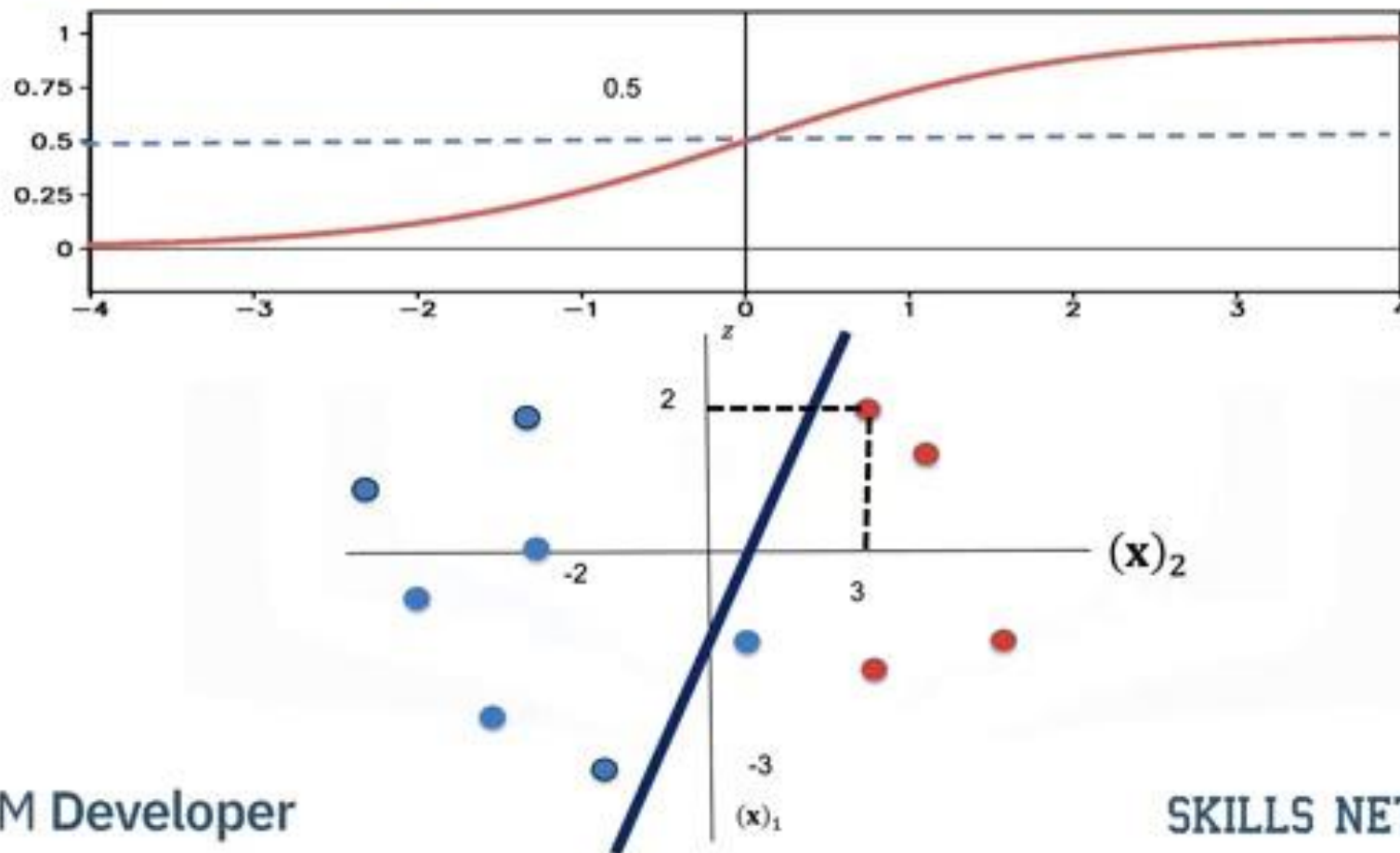
Logistic Regression



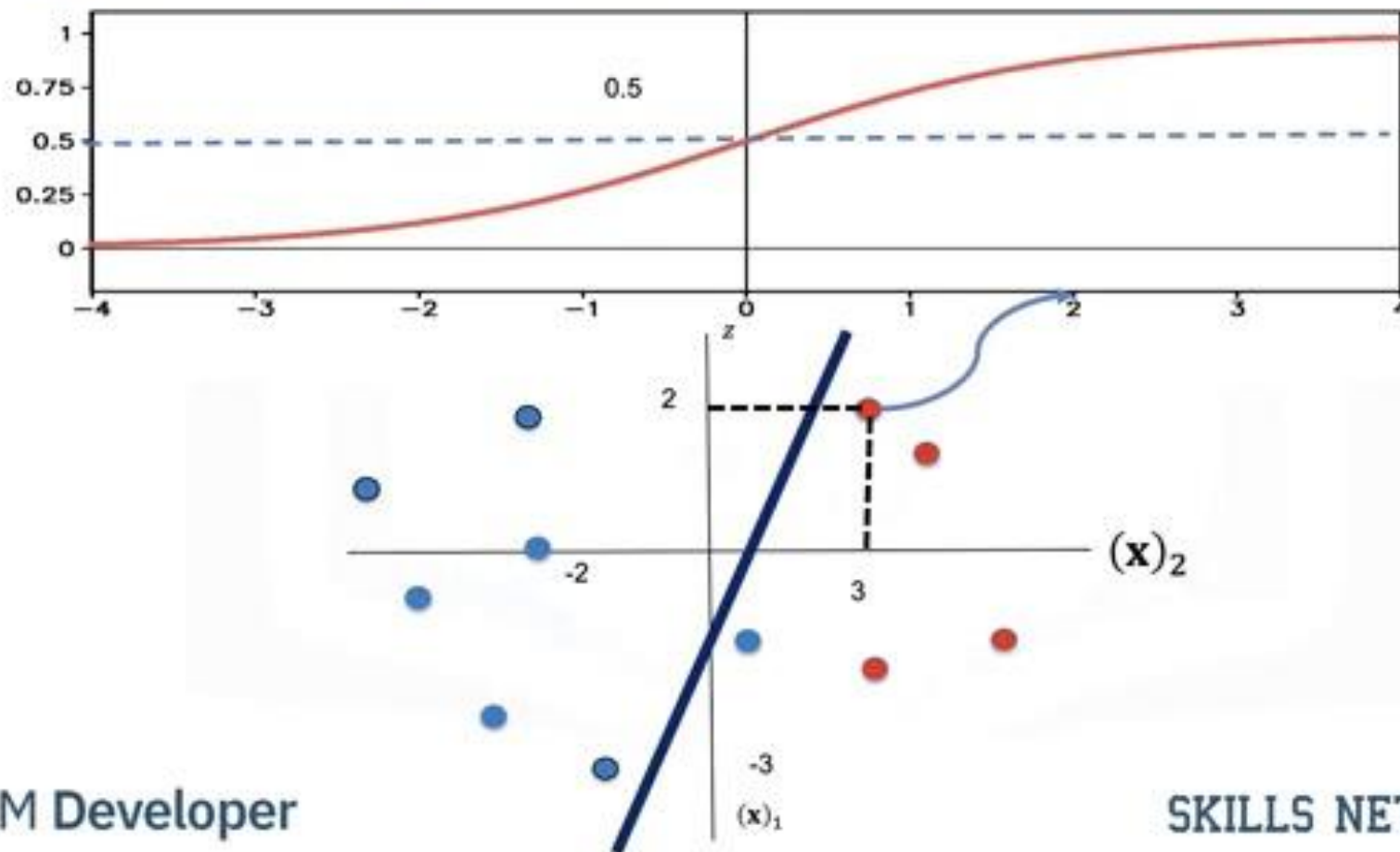
Logistic Regression



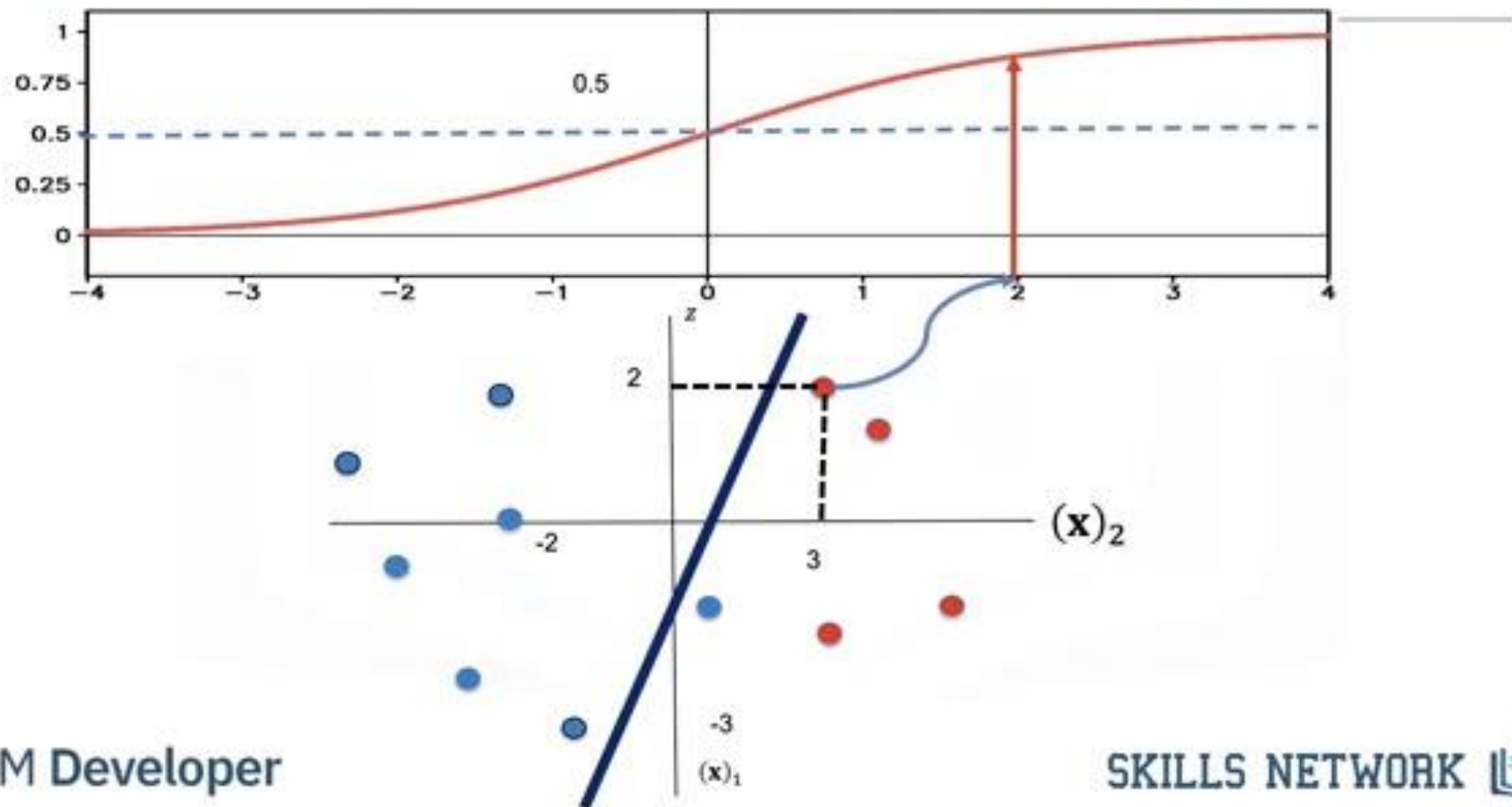
Logistic Regression



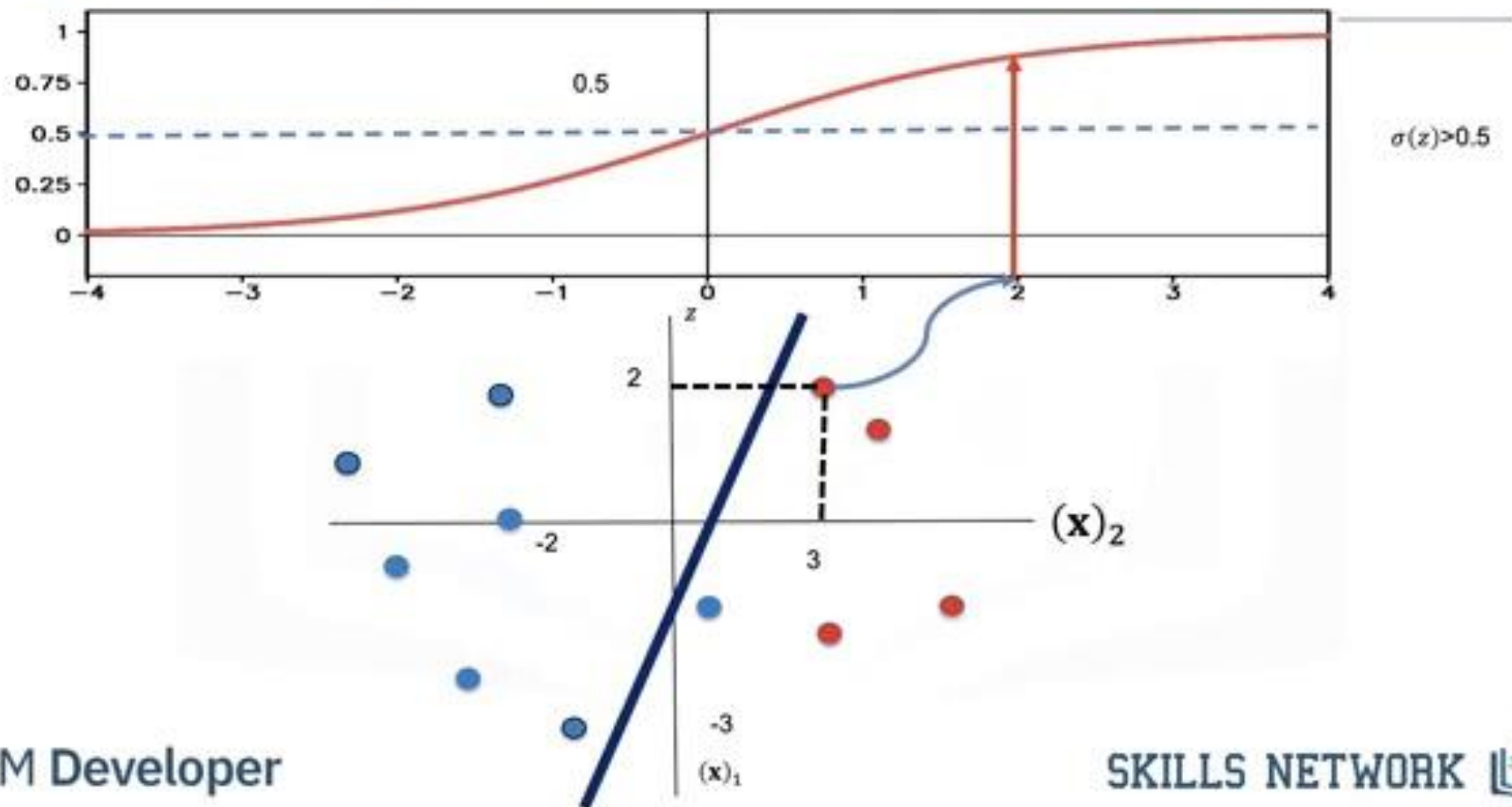
Logistic Regression



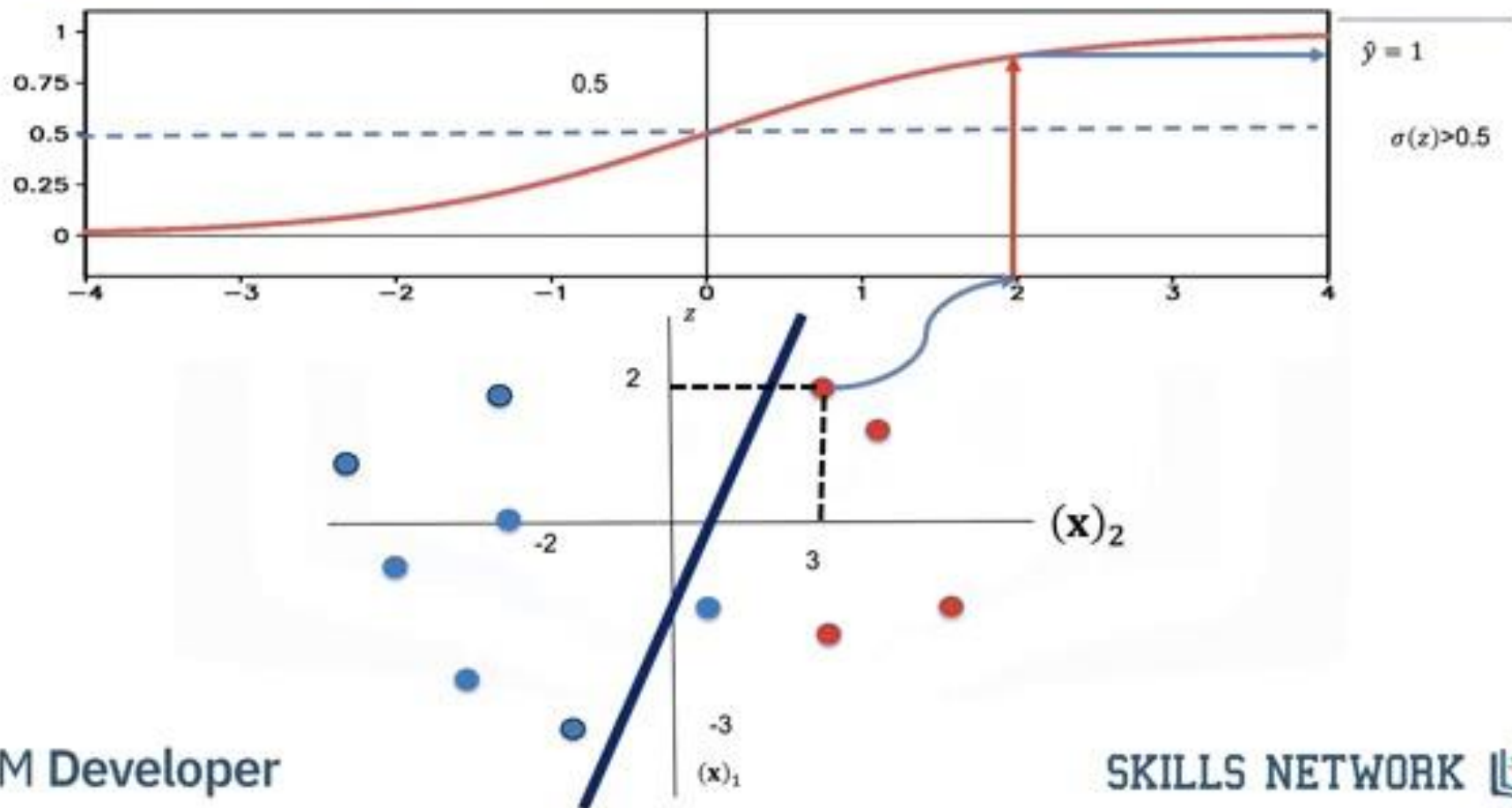
Logistic Regression



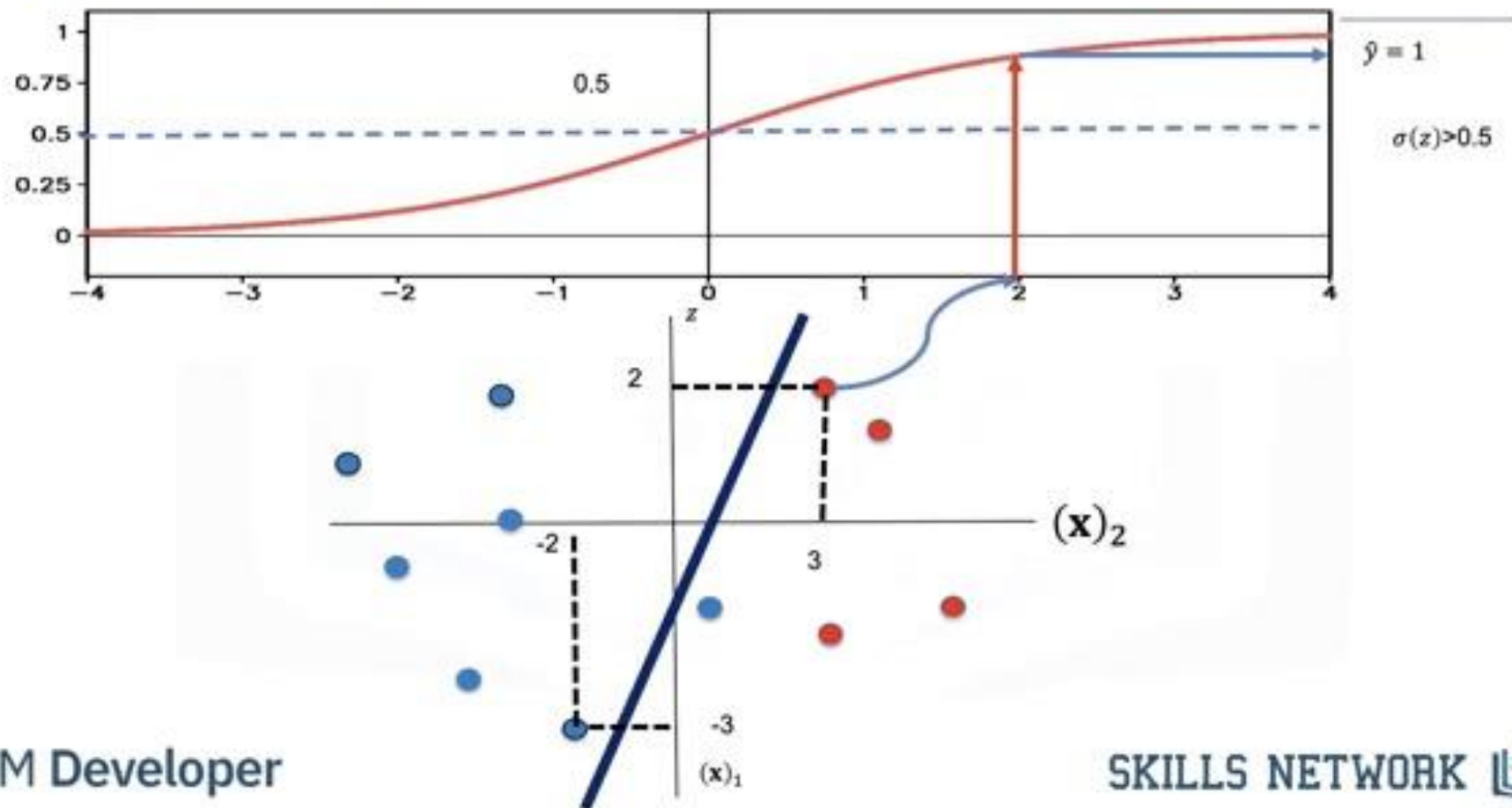
Logistic Regression



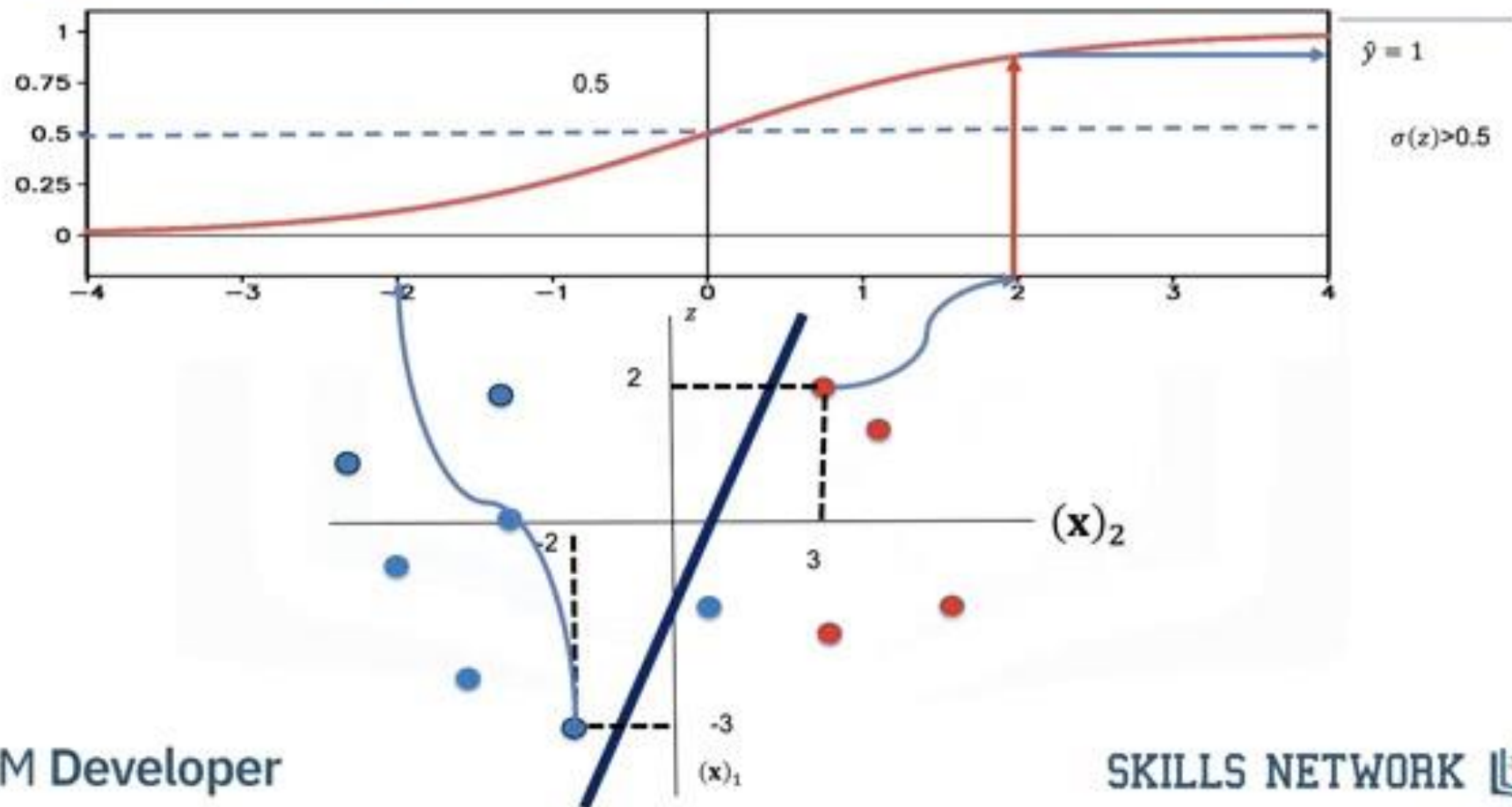
Logistic Regression



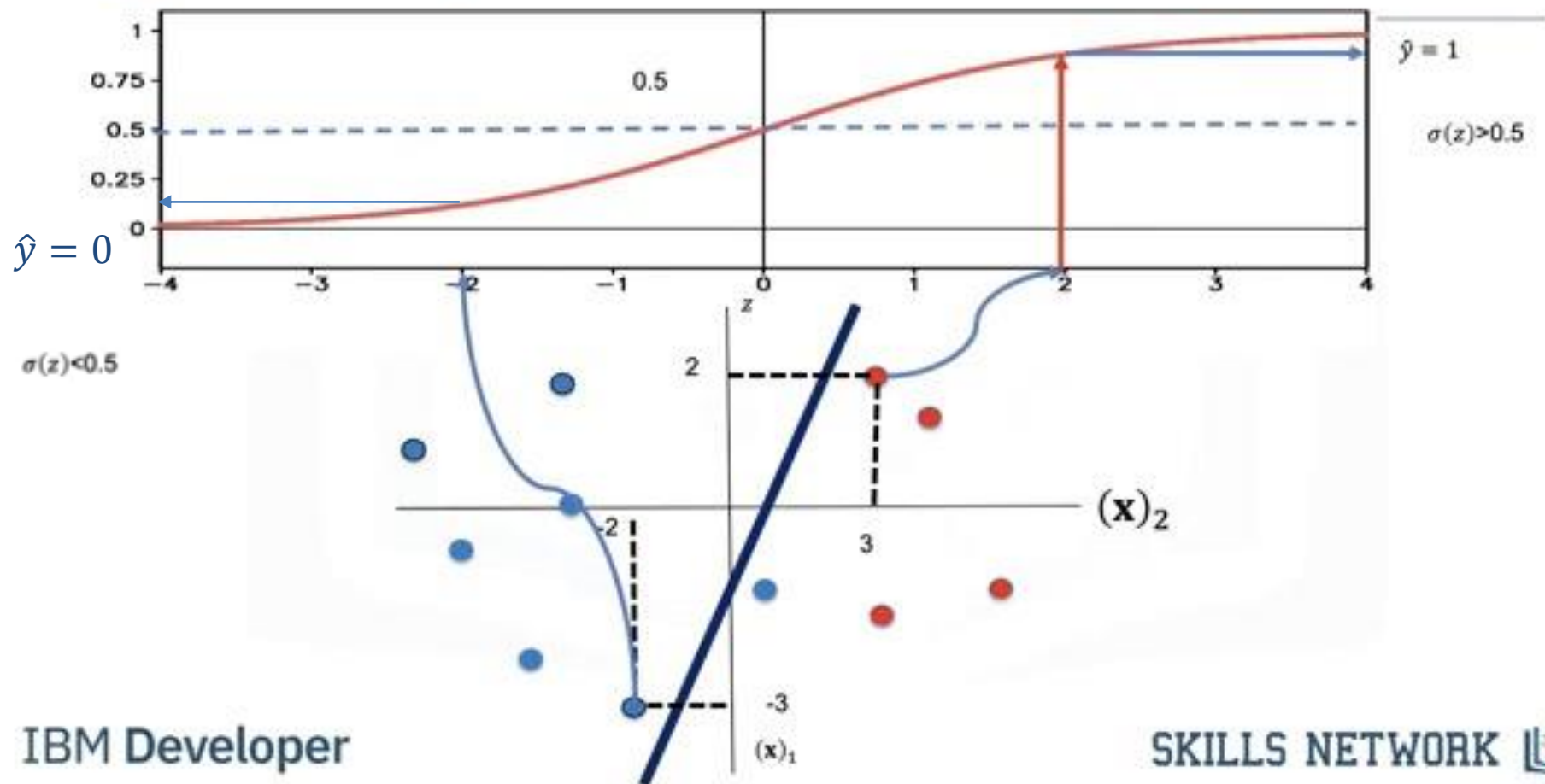
Logistic Regression



Logistic Regression



Logistic Regression



$$\overset{\sigma}{p}(\text{Image of a Dog}) = 0.92 \text{ Probability this image is a Dog}$$



σ
 $p($



)

$= 0.92$ Probability this
image is a Dog

$1 - p($




)

$= 0.08$ Probability this
image is a Cat


$p($  $)$

$1 - p($  $)$

$$p(\text{) = 0.10$$

Probability this image is a Dog

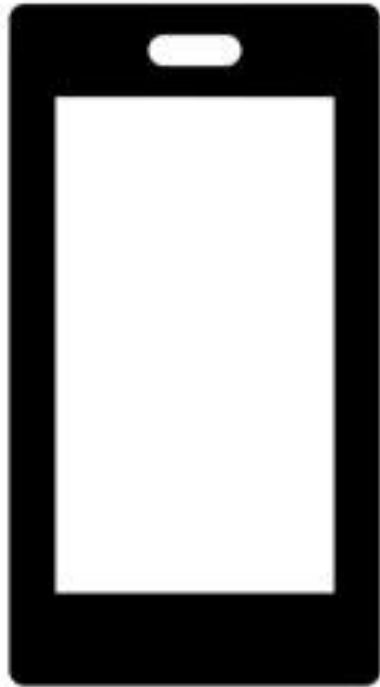
$$1 - p(\text{)$$

$$p(\text{) = 0.10$$

Probability this image is a Dog

$$1 - p(\text{) = 0.9$$

Probability this image is a Cat





$$\hat{y} = f(\text{image of cat})$$

$$\hat{y} = 0$$

- https://scikit-learn.org/stable/modules/generated/sklearn.linear_model.LogisticRegression.html



Examples

```
>>> from sklearn.datasets import load_iris
>>> from sklearn.linear_model import LogisticRegression
>>> X, y = load_iris(return_X_y=True)
>>> clf = LogisticRegression(random_state=0).fit(X, y)
>>> clf.predict(X[:2, :])
array([0, 0])
>>> clf.predict_proba(X[:2, :])
array([[9.8...e-01, 1.8...e-02, 1.4...e-08],
       [9.7...e-01, 2.8...e-02, ...e-08]])
>>> clf.score(X, y)
0.97...
```