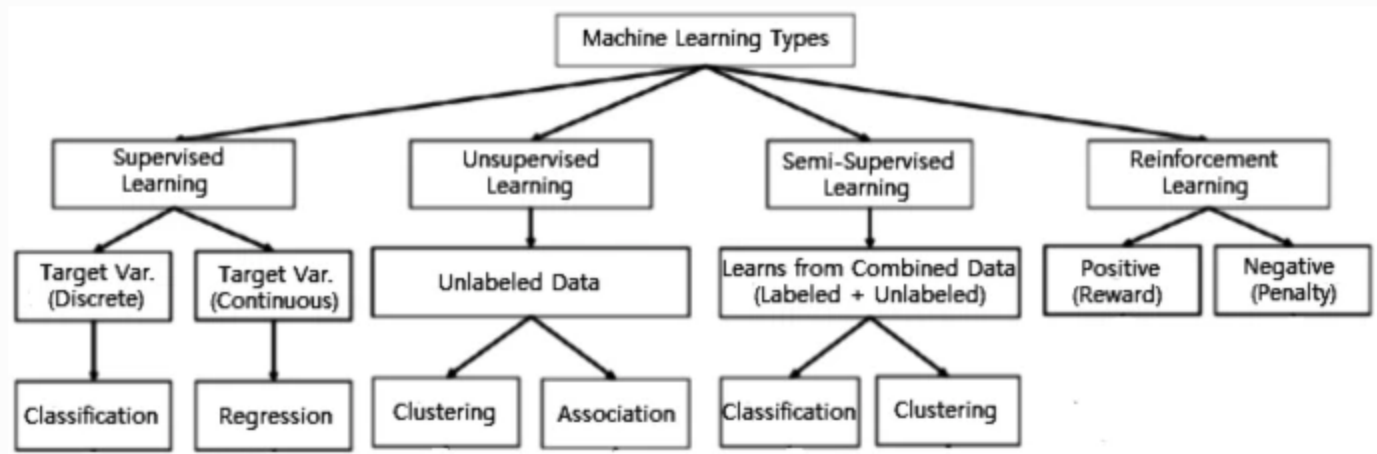


**Fig. 2**



Various types of machine learning techniques

Ref: <https://link.springer.com/article/10.1007/s42979-021-00592-x>

## What are best practices for ML?

### 1. Starting with an Interpretable Model

Keep the first model simple and get the infrastructure right far of any complications

### 2. Use Checkpoints:

A checkpoint is an intermediate dump of a model's internal state (parameters and hyperparameters), that will give high performance and less training time

### 3. Performance over fancy metric

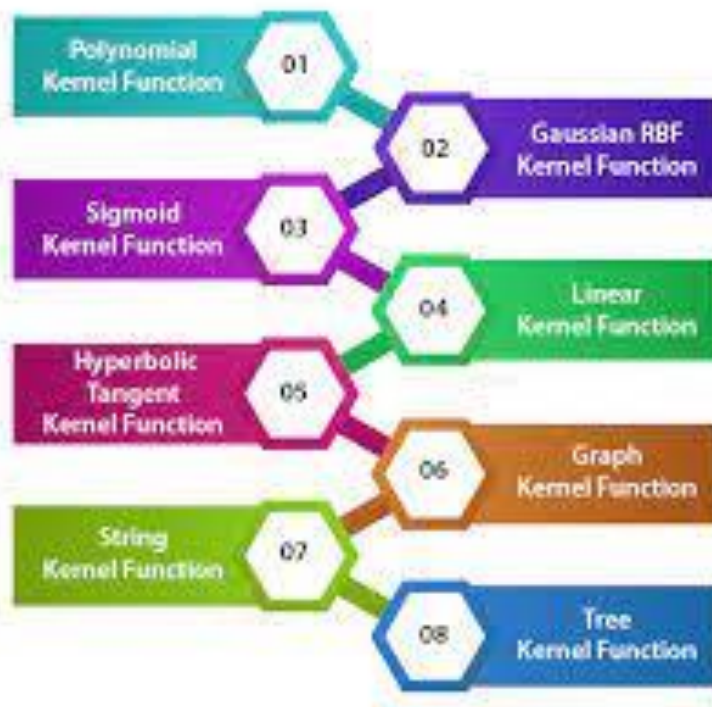
if there is some change that improves log loss but degrades the performance of the system, look for another feature.

### 4. Production Data to Training Data

store every new sample that comes from the serving model and then uses it for training.

Ref: <https://rubikscore.net/2021/10/18/18-machine-learning-best-practices/>

## ***Types of Kernel Functions***



### **linear Kernel**

It proves to be the best function when there are lots of features. The linear kernel is mostly preferred for text-classification problems as most of these kinds of classification problems can be linearly separated.

Linear kernel functions are faster than other functions.

$$F(x, x_j) = \text{sum}(x \cdot x_j)$$

Here,  $x, x_j$  represents the data you're trying to classify.

## Polynomial Kernel

A General representation of linear kernel. It is not as preferred as other kernel functions as it is less efficient and accurate.

$$F(x, x_j) = (x \cdot x_j + 1)^d$$

Here '.' shows the **dot product** of both the values, and **d** denotes the degree.

$F(x, x_j)$  representing the **decision boundary** to separate the given classes.

## Gaussian Radial Basis Function (RBF)

It is one of the most preferred and used kernel functions in svm. It is usually chosen for **non-linear data**. It helps to make proper separation when there is no prior knowledge of data.

$$F(x, x_j) = \exp(-\gamma * ||x - x_j||^2)$$

The value of gamma varies from **0 to 1**. You have to manually provide the value of gamma in the code. The most preferred value for **gamma is 0.1**.

## Sigmoid Kernel

It is mostly preferred for **neural networks**. This kernel function is similar to a two-layer perceptron model of the neural network, which **works as an activation function** for neurons.

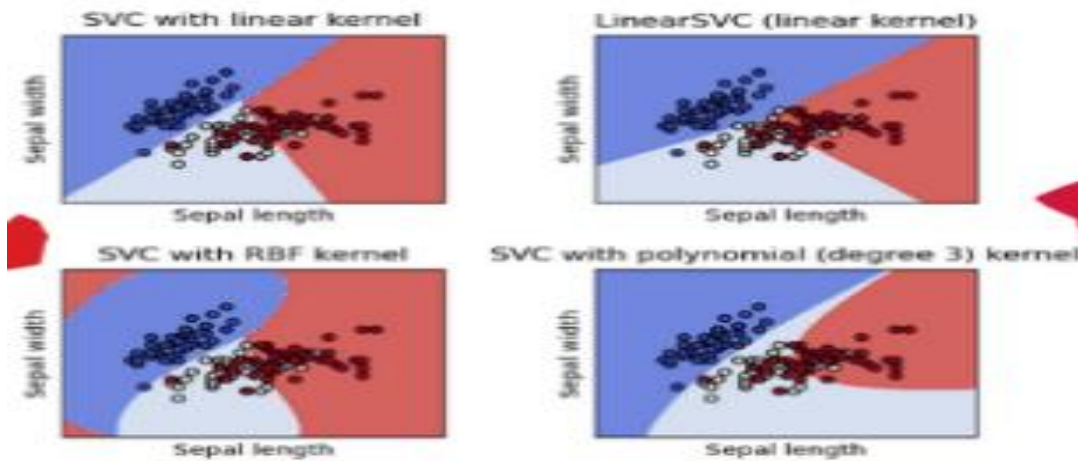
It can be shown as,

$$F(x, x_j) = \tanh(\alpha x_j y + c)$$

## Gaussian Kernel

It is a commonly used kernel. It is used when there is no prior knowledge of a given dataset.

$$k(x, y) = \exp\left(-\frac{\|x - y\|^2}{2\sigma^2}\right)$$



Ref : <https://dataaspirant.com/svm-kernels/>