

< What is a Gaussian Process >

: A Gaussian process is a non parametric model used for regression. It doesn't
(e.g. a b in $y = ax + b$)

a fixed form. Instead it defines a distribution over functions.

"I don't know the exact function that maps input to output, but I'll assume that any finite collection is jointly Gaussian distributed."

* Key intuition

Suppose you're trying to model an unknown function $y = f(x)$

You only have a few observed input-output pairs.

$x_1, x_2, \dots, x_n \quad y_1, y_2, \dots, y_n$

A Gaussian process gives a distribution over all possible functions that

Fit the known data points. \Rightarrow Follow a certain smoothness or structure controlled by a kernel func.

* Core Components.

1. Mean functions.

$m(x) = \mathbb{E}(f(x)) = 0$ usually assumed zero.

2. Covariance Function (Kernel)

$$K(x, x') = \text{Cov}(f(x), f(x'))$$

If x, x' are close $K(x, x')$ is large \rightarrow outputs are highly correlated (smoothness)

* How GP regression Works.

1. Given training data $X_{\text{train}} \quad Y_{\text{train}}$.

2. Want to predict at test points X_{test}

3.
$$\begin{bmatrix} y_{\text{train}} \\ y_{\text{test}} \end{bmatrix} \sim \mathcal{N} \left(0, \begin{bmatrix} K(X_{\text{train}}, K_{\text{train}}) & K(X_{\text{train}}, X_{\text{test}}) \\ K(X_{\text{test}}, X_{\text{train}}) & K(X_{\text{test}}, X_{\text{test}}) \end{bmatrix} \right)$$

μ : predicted value / σ^2 : uncertainty variance at that point.

* Advantages

Uncertainty Quantification: not only predictions but also their confidence intervals.

Non parametric: No need to choose a specific model shape.

Small dataset: Powerful with few data points.