

Lecture 21: Gaussian process regression

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The effect of the covariance function function

The effect of the covariance function - The Squared exponential

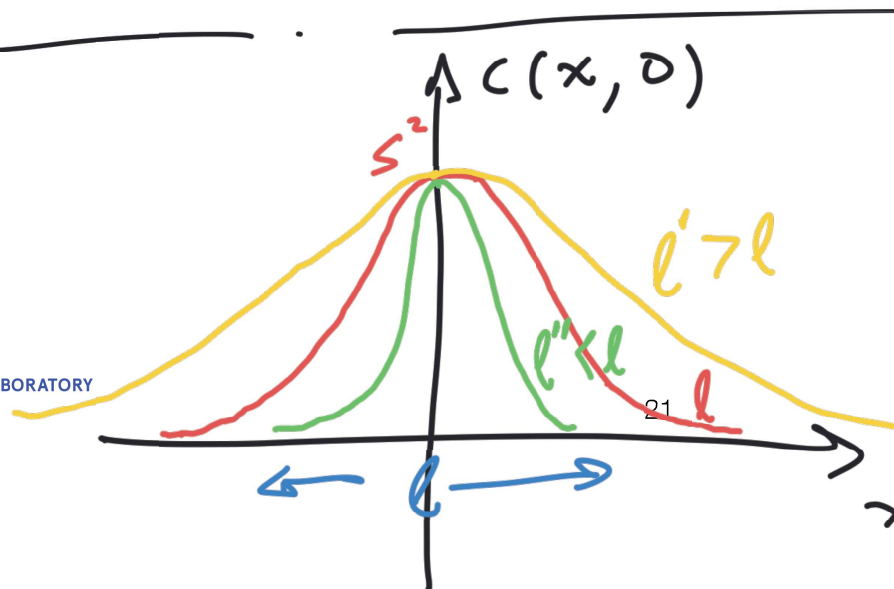
$d=1$: $x \in \mathbb{R}$

$$c(x, x') = \underbrace{s^2}_{\text{signal variance}} \exp \left\{ - \frac{(x - x')^2}{2 \underbrace{\ell^2}_{\text{length scale}}} \right\}$$

←

$d > 1$:

$$c(x, x') = s^2 \exp \left\{ - \sum_{i=1}^d \frac{(x_i - x'_i)^2}{2 \ell_i^2} \right\}$$



$$c(0, 0) = s^2$$

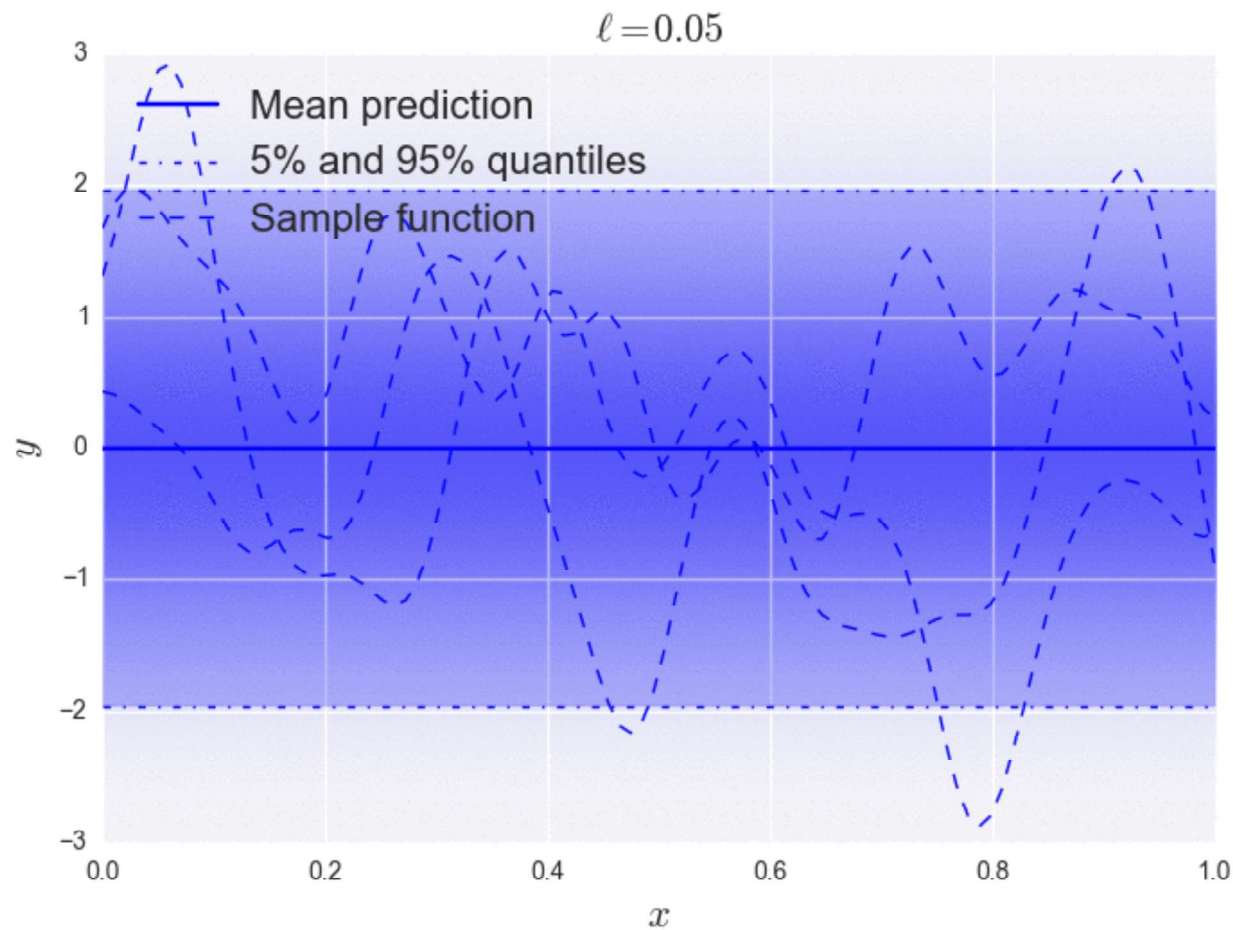
$$c(x, 0) \rightarrow 0, \text{ as } x \rightarrow \pm \infty$$

$$s^2: V[f(x)] \stackrel{\text{def.}}{=} c(x, x) \stackrel{\text{SE}}{=} s^2$$

$$x \ell: \ell \uparrow \quad c(x, x') \uparrow$$



Changing the length scale



function - Regularity

$$f(\cdot) \sim GP(0, C(\cdot, \cdot))$$

Thm: Regularity of samples from the GP \Leftrightarrow cov. fun. C is the same as the regularity of $g(x) = C(x, x)$.

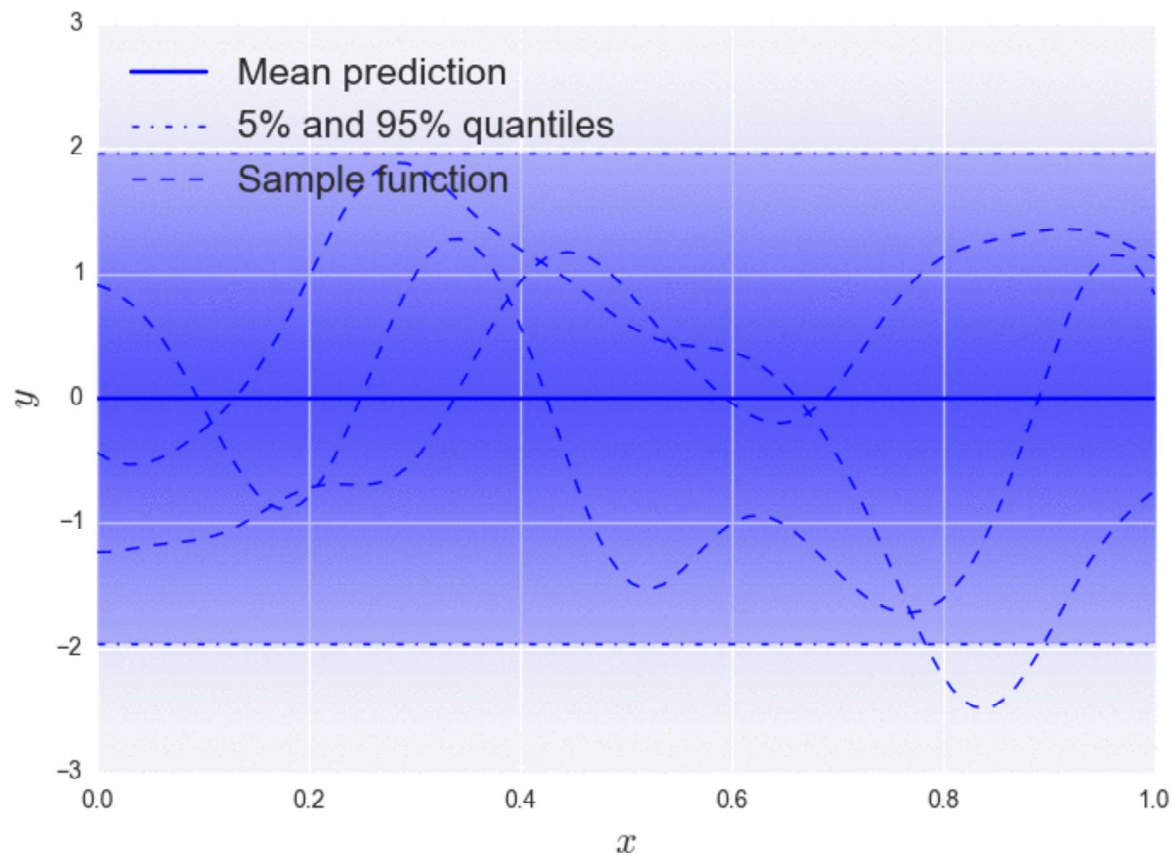
- f is continuous at x if $g(x) = C(x, x)$ is cont. at x .
- $\frac{\partial f}{\partial x_j}$ is continuous at x if $\frac{\partial^2 C(x, x)}{\partial x_j \partial x_j'}$ is cont. at x .
- $\frac{\partial^2 f}{\partial x_j \partial x_k}$ is continuous at x if $\frac{\partial^4 C(x, x)}{\partial x_j \partial x_k \partial x_j' \partial x_k'}$ is cont. at x .

SE cov. fun. ²³

infinitely diff. $\Rightarrow f$ int. diff.

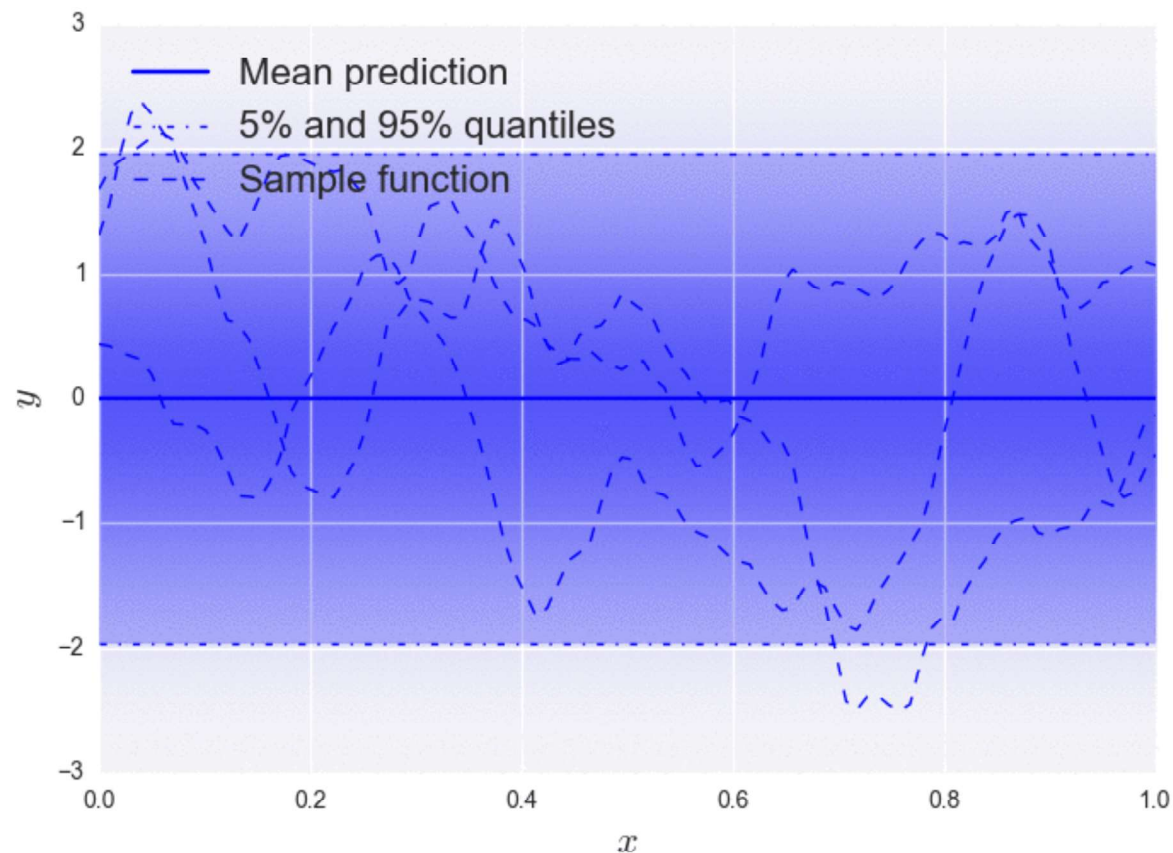
The samples are as smooth as the covariance

Infinitely smooth SE covariance



The samples are as smooth as the covariance

Matern 2-3, 2 times differentiable



The samples are as smooth as the covariance

Exponential, continuous, nowhere differentiable

