#### Gaussian Processes Visual Tool

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#### Introduction



GP Visual Tool is called to be a system for the interactive modeling, fitting and interpreting of Gaussian processes.

TODO: Overall image

It allows the user to rigorously specify a model by choosing different sets of hyperparameters.

### What is a Gaussian process?



Typically, Gaussian processes (GPs) can be seen as a generalization of the Bayesian Regression. Suppose we want to model a  $f: \mathbb{R} \to \mathbb{R}$  as a GP. We would define it as following:

$$\mathbf{y} \mid (\mathbf{f}, \mathbf{x}) \sim \mathcal{N}(\mathbf{f}, \sigma^2 \mathbf{I}),$$
  
 $\mathbf{f} \mid \mathbf{x} \sim \mathcal{GP}(m, k) \equiv \mathcal{N}(m(\mathbf{x}), k(\mathbf{x}, \mathbf{x})),$ 

for  $\mathbf{x} = x_1, \dots, x_N$ ,  $\mathbf{y} = y_1, \dots, y_N$  and  $\mathbf{f} = f(x_1), \dots, f(x_N)$ . We often refer to m as the mean function (usually  $m(x_i) = 0$ ) and k as the kernel function. Our goal is to find the *posterior* distribution of f.

### Means and Kernels



# Finding the *posterior* distribution



# Some approaches to get optimal hyperparams



# Applications of GPs



### Life could be a dream.



# Developing a visual tool



# Objectives and interactions.



# Let's start sampling



# Some marginals are not that bad<sup>1</sup>



<sup>&</sup>lt;sup>1</sup>Marginal distributions :D

# Beyond choosing a kernel



# Math is actually important



### Customization is allowed



# Time to play!





# Thanks!