Automatic Construction and Natural-Language Description of Nonparametric Regression Models







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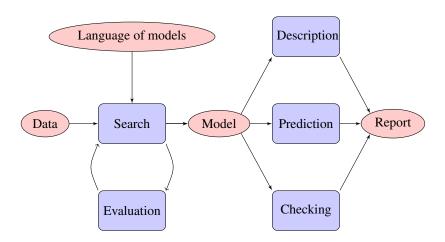




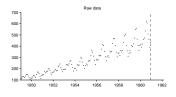
Joshua Tenenbaum², Zoubin Ghahramani¹

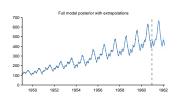
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A SYSTEM FOR AUTOMATIC DATA ANALYSIS

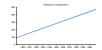


AN ENTIRELY AUTOMATIC ANALYSIS

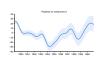




Four additive components have been identified in the data



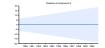
A linearly increasing function



A smooth function



An approximately periodic function with a period of 1.0 years with linearly increasing amplitude



Uncorelated noise with linearly increasing standard deviation

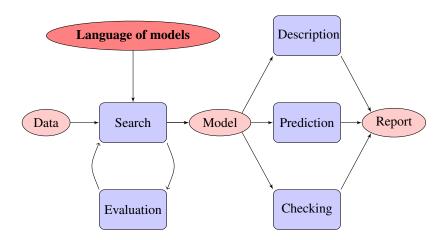
NATURAL LANGUAGE DESCRIPTIONS OF MODELS

Compositionally constructed statistical models

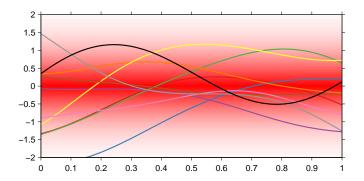


Compositionally constructed natural-language descriptions

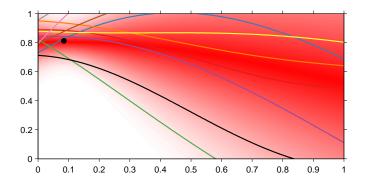
DEFINING A LANGUAGE OF MODELS



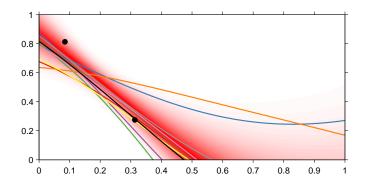
- ▶ Define probability distributions on functions
- ► Used to perform Bayesian (nonlinear) regression



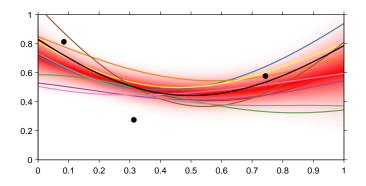
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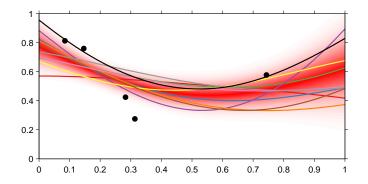
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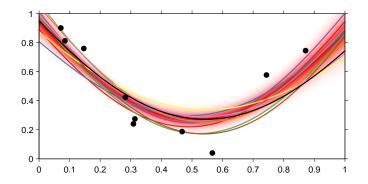
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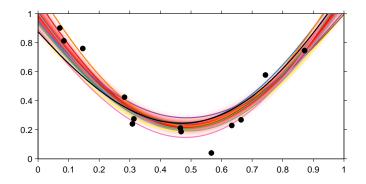
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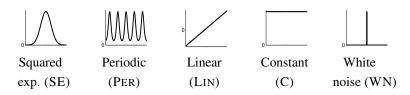


- ► Define probability distributions on functions
- ▶ Used to perform Bayesian (nonlinear) regression

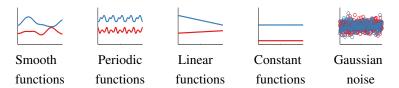


THE ATOMS OF OUR LANGUAGE

Five base kernels...

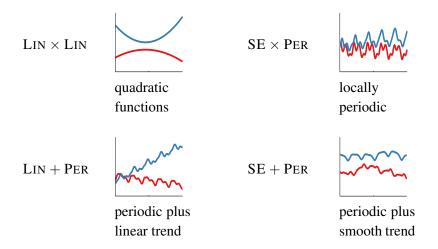


...encoding for the following types of functions

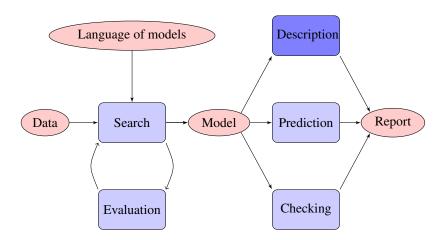


THE COMPOSITION RULES OF OUR LANGUAGE

► Two main operations: addition, multiplication



AUTOMATIC TRANSLATION OF MODELS

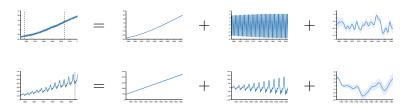


SUMS OF KERNELS ARE SUMS OF FUNCTIONS

If $f_1 \sim GP(0, k_1)$ and independently $f_2 \sim GP(0, k_2)$ then

$$f_1 + f_2 \sim \text{GP}(0, \frac{k_1}{k_1} + \frac{k_2}{k_2})$$

e.g.



We can therefore describe each component separately



- ▶ Properties of individual kernels well understood
- ► Can be described with standard noun phrase



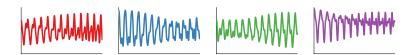






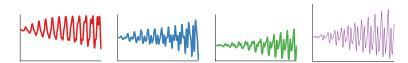
$$\underbrace{SE}_{\text{approximately}} \times \underbrace{PER}_{\text{periodic function}}$$

- ► Multiplying by each kernel has a consistent effect
- ► Can be described with consistent adjectives / modifiers



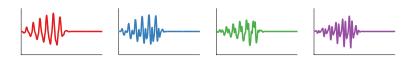
$$\underbrace{SE}_{approximately} \times \underbrace{PER}_{periodic function} \times \underbrace{LIN}_{with linearly growing amplitude}$$

- ▶ Multiplying by each kernel has a consistent effect
- ► Can be described with consistent adjectives / modifiers



$$\underbrace{\text{SE}}_{\text{approximately}} \times \underbrace{\text{PER}}_{\text{periodic function}} \times \underbrace{\text{LIN}}_{\text{with linearly growing amplitude}} \times \underbrace{\boldsymbol{\sigma}}_{\text{until 1700}}$$

- ► Multiplying by each kernel has a consistent effect
- ► Can be described with consistent adjectives / modifiers



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James Robert Lloyd 15/1.