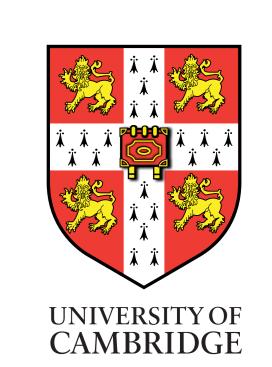


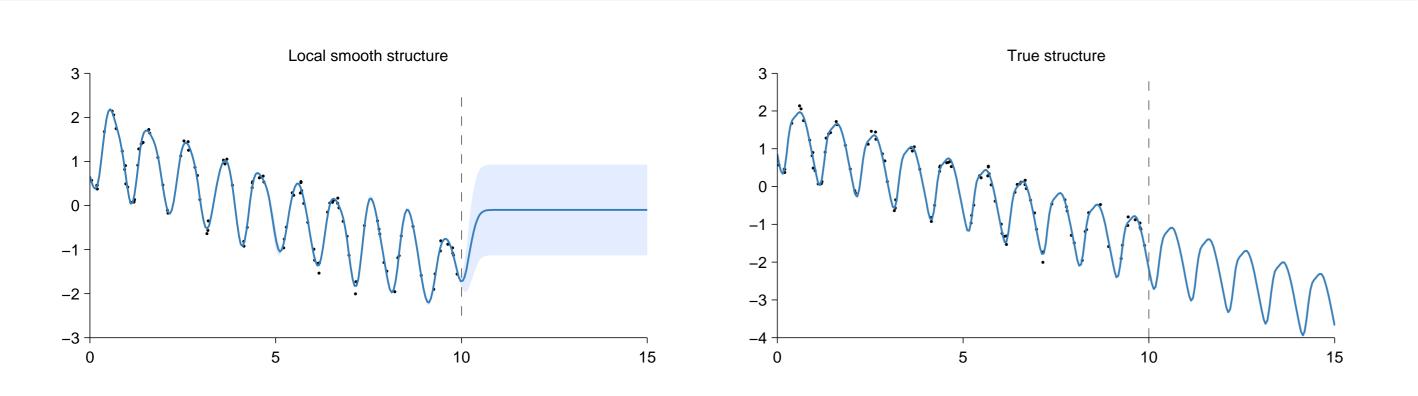
Technology

Structure Discovery in Nonparametric Regression through Compositional Kernel Search



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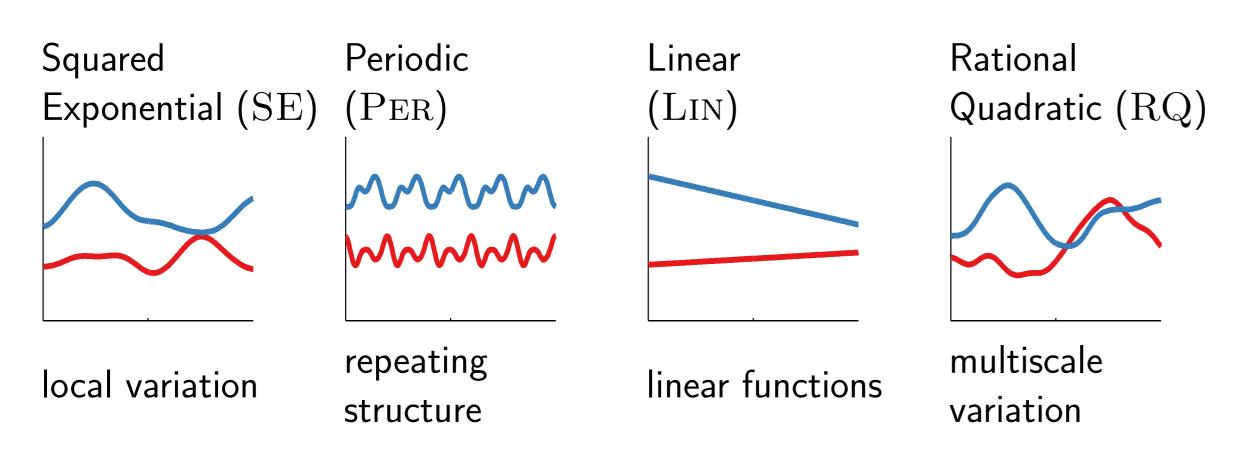
Identifying structure is crucial for extrapolation



- Traditionally, a statistician proposes an appropriate model for the type of structures present
- Automatic model selection techniques already exist, typically choosing between a finite or restricted set of models
- Instead, we automate statistical model construction

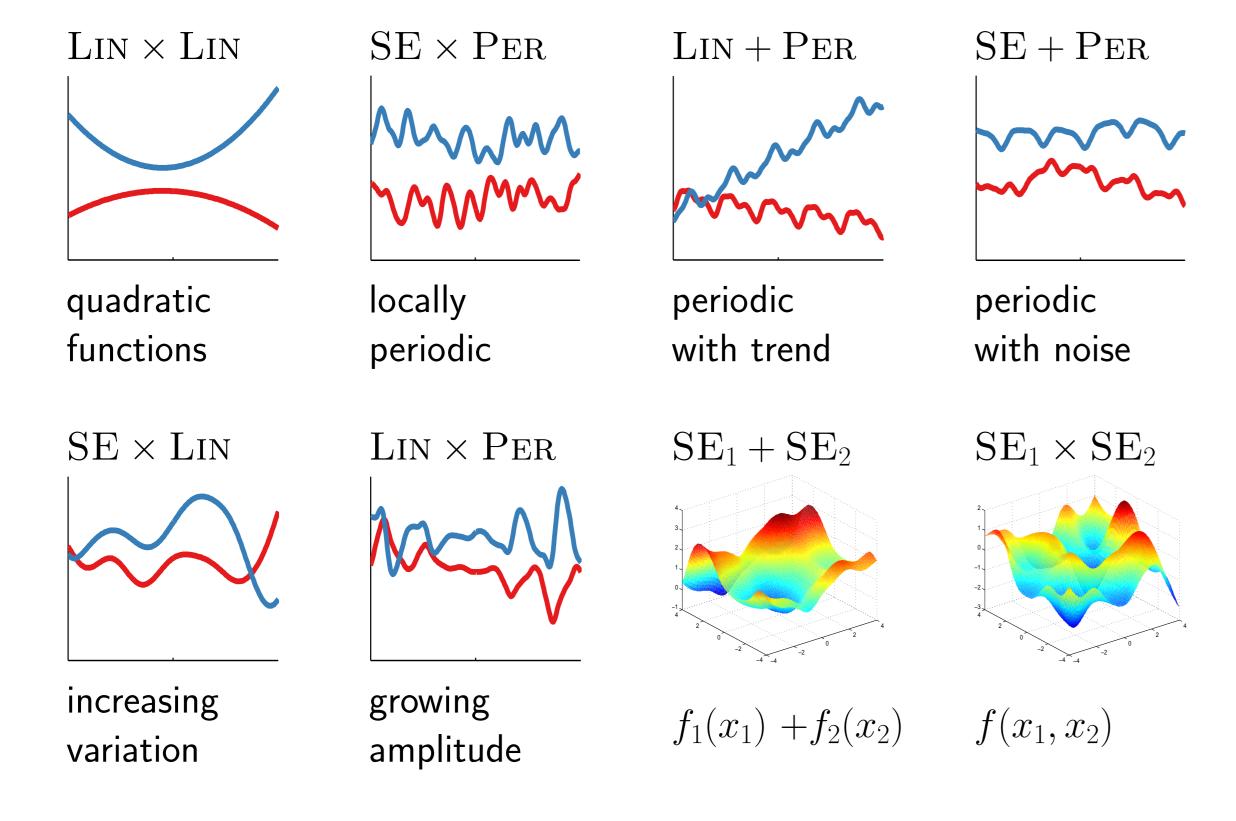
Gaussian processes model structure through kernels

- The kernel specifies which structures are likely under the GP prior, which in turn determines the generalization properties of the model.
- Below we list standard base kernels, and draws from the corresponding GP priors:



Kernels can be composed...

• Constructing appropriate composite kernels has previously been done by experts



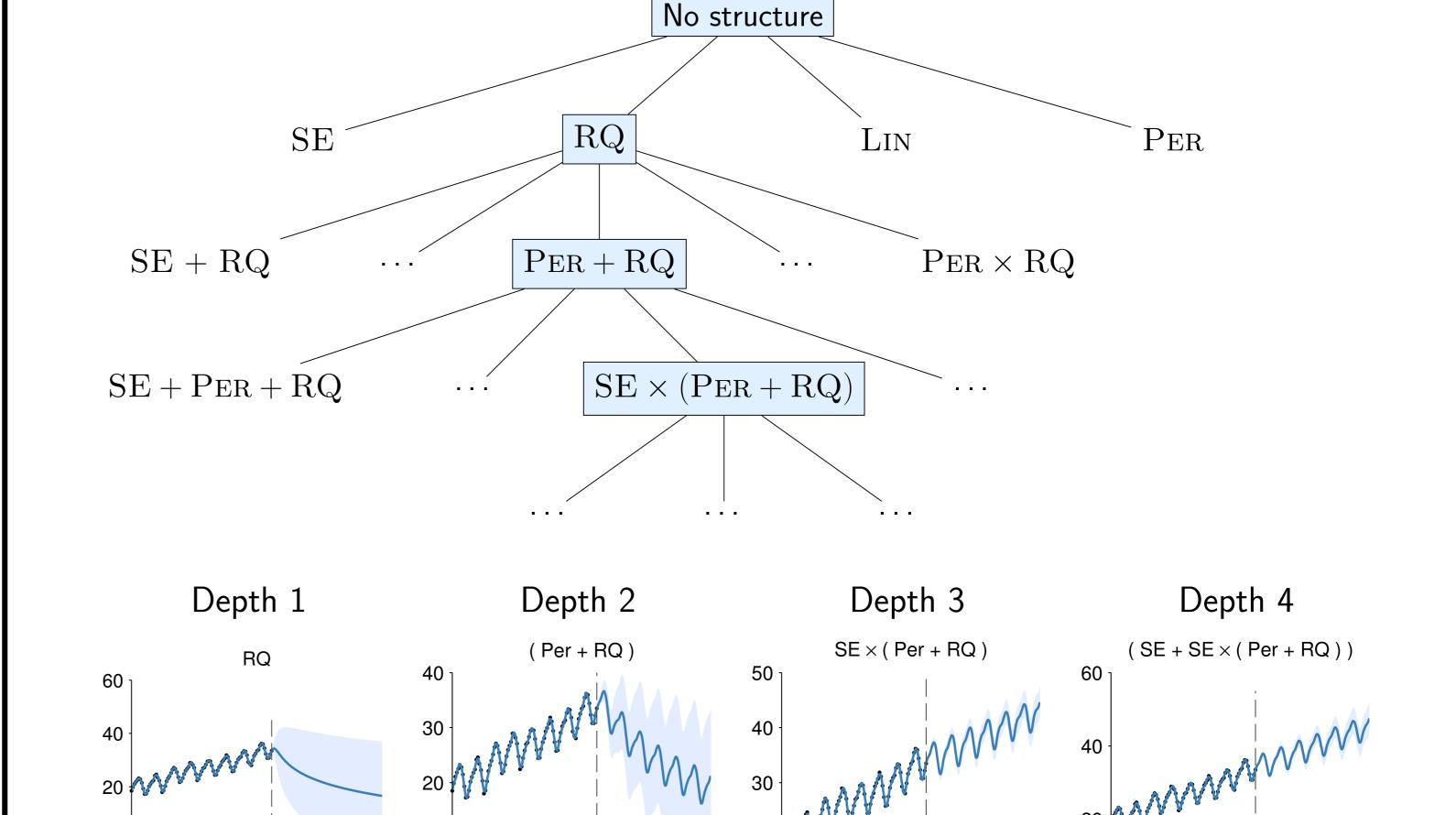
... defining a rich, open-ended set of models...

ullet We consider all algebraic expressions composed a small number of base kernels and the operations '+' and '×'

Special cases of our model

. . . which we search greedily

- We try all base kernels, selecting the one with the highest (approximate) marginal likelihood which balances data fit and model complexity
- The search continues by adding an extra term to the current best kernel, stopping when marginal likelihood no longer improves



Compound kernels are interpretable

2010

2000

2005

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• Compound kernels decompose functions into additive components

2000

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