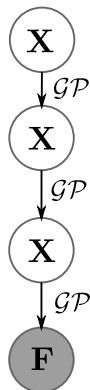


Deep Gaussian Processes

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Refer to the slide of Andreas Damianou at GPSS

Deep Gaussian processes



- Now recurse the stacked construction

$$f(\mathbf{x}) \rightarrow \text{GP}$$

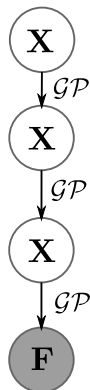
$$f(x(\mathbf{t})) \rightarrow \text{stacked GP}$$

$$f(x_2(\mathbf{x}_1)) \rightarrow \text{stacked GP}$$

$$f(x(x(x \cdots (\mathbf{x}_1)))) \rightarrow \text{deep GP}$$

- The variational approximation changes only a little
- Uncertainty modelled “everywhere”!

Deep Gaussian processes



- Now recurse the stacked construction

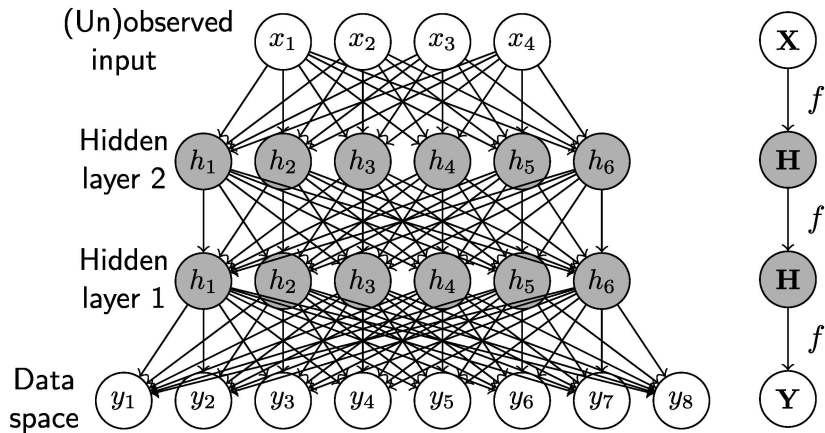
$$f(\mathbf{x}) \rightarrow \text{GP}$$

$$f(x(\mathbf{t})) \rightarrow \text{stacked GP}$$

$$f(x_2(\mathbf{x}_1)) \rightarrow \text{stacked GP}$$

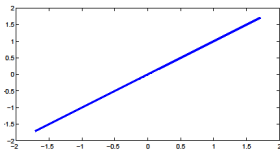
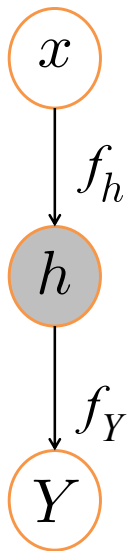
$$f(x(x(x \cdots (\mathbf{x}_1)))) \rightarrow \text{deep GP}$$

Deep learning

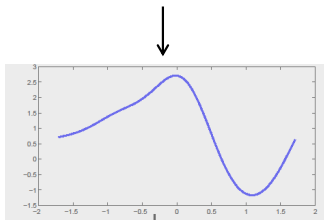


$$\mathbf{Y} = f(f(\cdots f(\mathbf{X})))$$

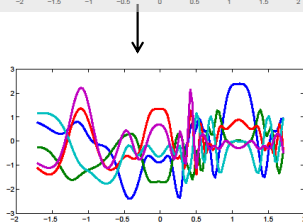
Sampling from a deep GP



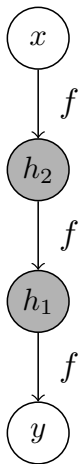
Input



Unobserved



Output



- ▶ Joint = $p(y|h_1)p(h_1|h_2)p(h_2|x)$
- ▶ MAP optimization is extremely problematic because:
 - Dimensionality of h s has to be decided a priori
 - Prone to overfitting, if h are treated as parameters
 - Deep structures are not supported by the model's objective but have to be forced [Lawrence & Moore '07]

Regularization solution: approximate Bayesian framework

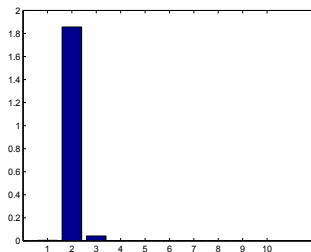
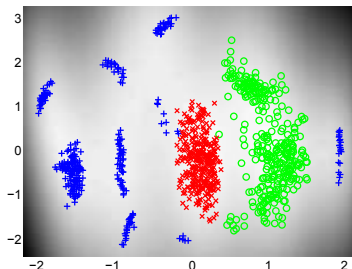
- ▶ Analytic variational bound $\mathcal{F} \leq p(y|x)$
 - Extend Titsias' method for *variational learning of inducing variables in Sparse GPs*.
 - *Approximately* marginalise out h
- ▶ Automatic structure discovery (nodes, connections, layers)
 - Use the Automatic / Manifold Relevance Determination trick
- ▶ ...

Automatic dimensionality detection

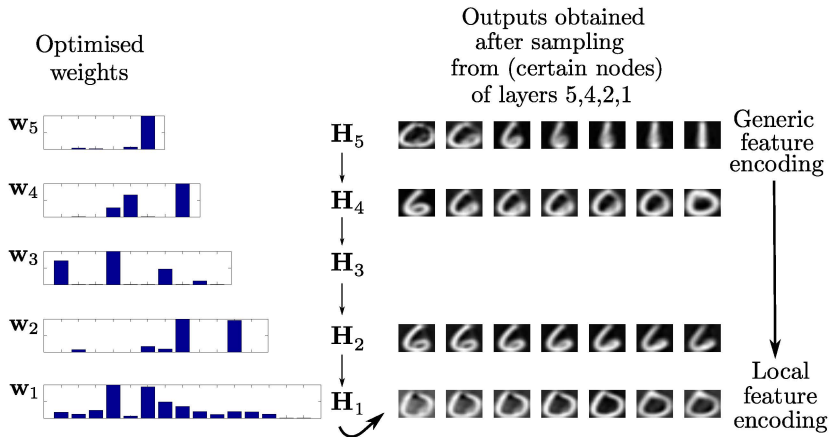
- ▶ Achieved by employing *automatic relevance determination* (ARD) priors for the mapping f .
- ▶ $f \sim \mathcal{GP}(\mathbf{0}, k_f)$ with:

$$k_f(\mathbf{x}_i, \mathbf{x}_j) = \sigma^2 \exp \left(-\frac{1}{2} \sum_{q=1}^Q w_q (x_{i,q} - x_{j,q})^2 \right)$$

- ▶ Example:



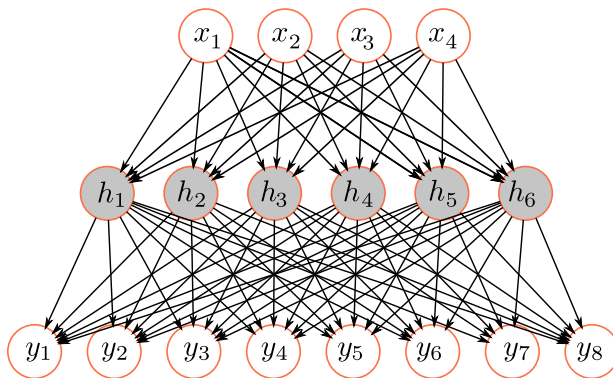
Deep GP: MNIST example



Automatic structure discovery

Tools:

- ▶ ARD: Eliminate unnecessary nodes/connections
- ▶ MRD: Conditional independencies
- ▶ Approximating evidence: Number of layers (?)



Automatic structure discovery

Tools:

- ▶ ARD: Eliminate unnecessary nodes/connections
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