

*Encode prior assumptions
(e.g. dynamics [5])
or place observed inputs
(supervised learning)*

$$\mathbf{X}_H = [\mathbf{x}_H^{(1)}, \dots, \mathbf{x}_H^{(Q_H)}]$$

f_{H-1}

\dots

f_1

$$\mathbf{X}_1 = [\mathbf{x}_1^{(1)}, \dots, \mathbf{x}_{(1)}^{(Q_1)}]$$

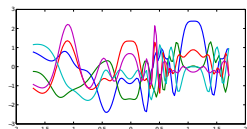
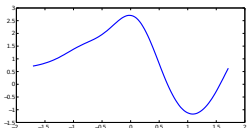
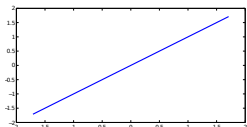
f_0

$$\mathbf{X}_0 = [\mathbf{x}_0^{(1)}, \dots, \mathbf{x}_0^{(Q_0)}]$$

*Hidden variables
(\mathbf{X}_H can also
be observed)*

*Single layer:
equivalent to
the GP-LVM [1]*

Observed outputs



$$f_h \sim \mathcal{GP}(\mathbf{0}, k_h(\mathbf{X}_h, \mathbf{X}_h))$$