

Actually Sparse Variational Gaussian Processes

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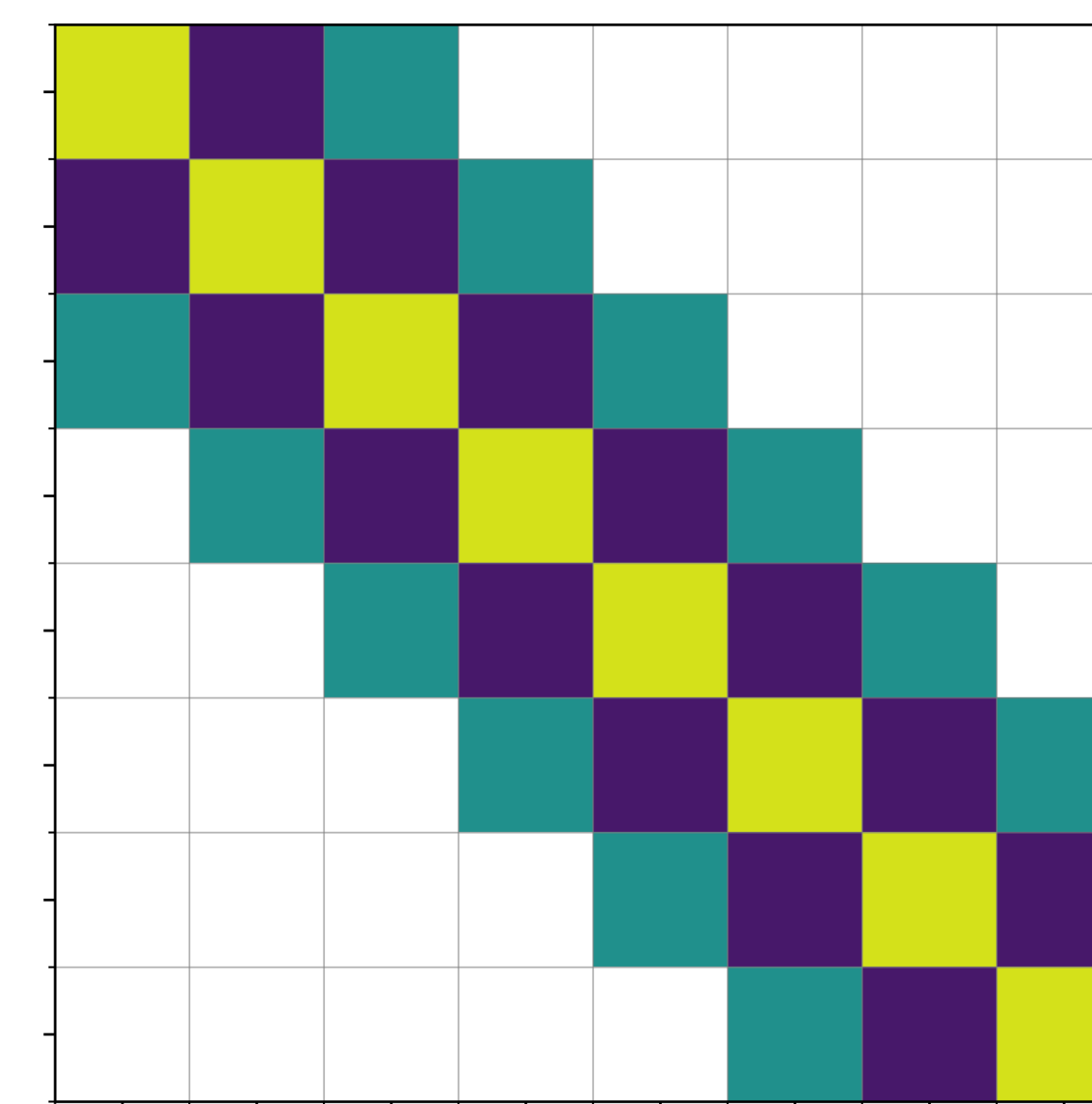
Contribution

- In sparse variational GP inference, **precomputation** (as done in VFF/VISH) is faster than **stochastic optimization** in SVGP
- However, computing the ELBO with VFF requires the **Cholesky factor** of the **dense matrix** ($\mathbf{K}_{\mathbf{u}\mathbf{u}} - \sigma^{-2}\mathbf{K}_{\mathbf{u}\mathbf{f}}\mathbf{K}_{\mathbf{f}\mathbf{u}}$)
- We overcome this by using **B-spline features** to make this matrix **sparse**, reducing the computational complexity to **linear in the number of inducing points**

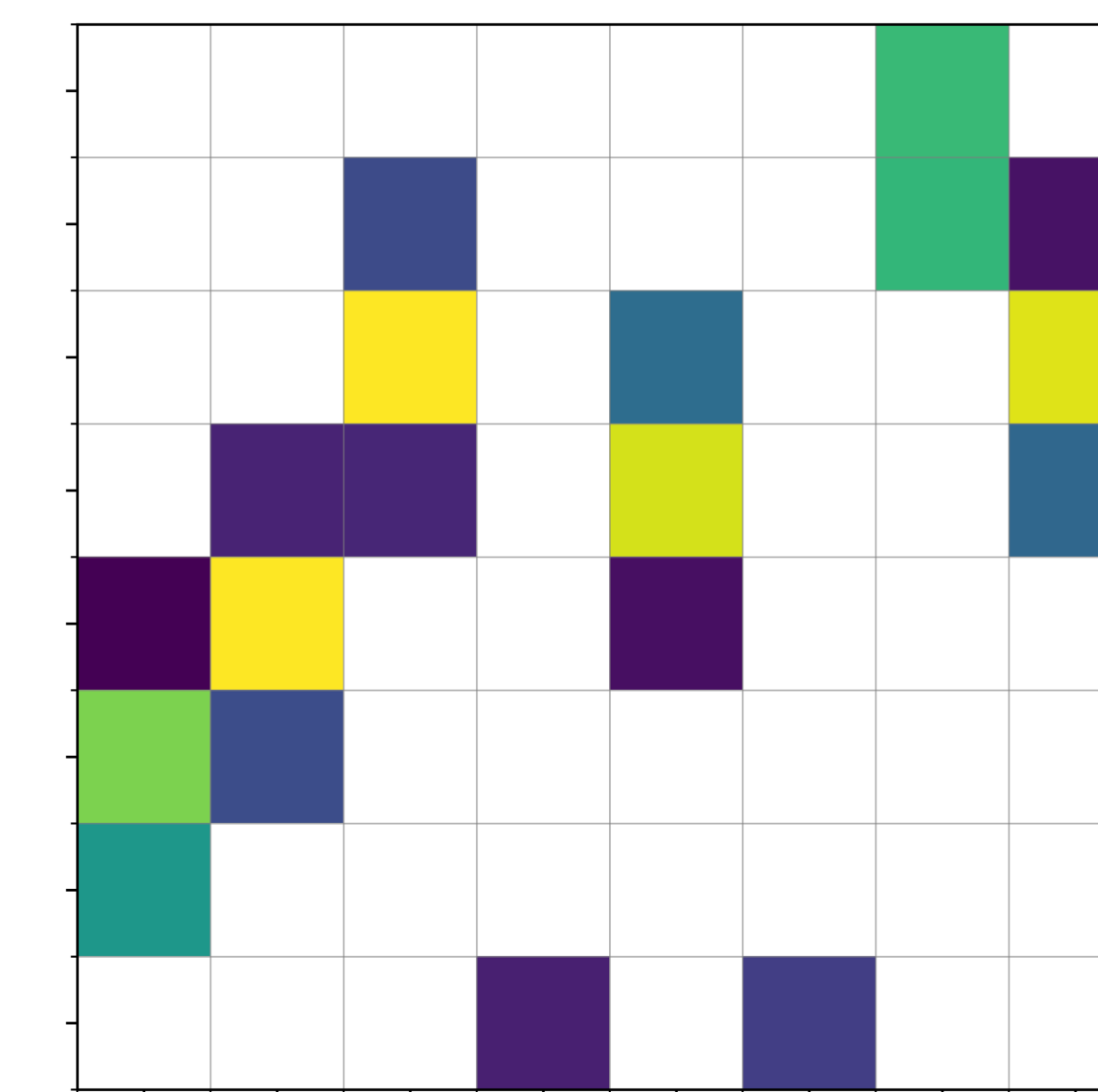
Algorithm	Pre-computation	Computational complexity	Storage
SGPR (Titsias, 2009)	\times	$\mathcal{O}(NM^2 + M^3)$	$\mathcal{O}(NM)$
SVGP (Hensman et al, 2013)	\times	$\mathcal{O}(N_b M^2 + M^3)$	$\mathcal{O}(M^2 + N_b M)$
VFF (Hensman et al, 2017)	$\mathcal{O}(NM^2)$	$\mathcal{O}(M^3)$	$\mathcal{O}(M^2 + NM)$
VISH (Dutordoir et al, 2020)	$\mathcal{O}(NM^2)$	$\mathcal{O}(M^3)$	$\mathcal{O}(M^2 + NM)$
AS-VGP (Ours)	$\mathcal{O}(N)$	$\mathcal{O}(Mk^2)$	$\mathcal{O}(Mk)$

Sparse Matrices

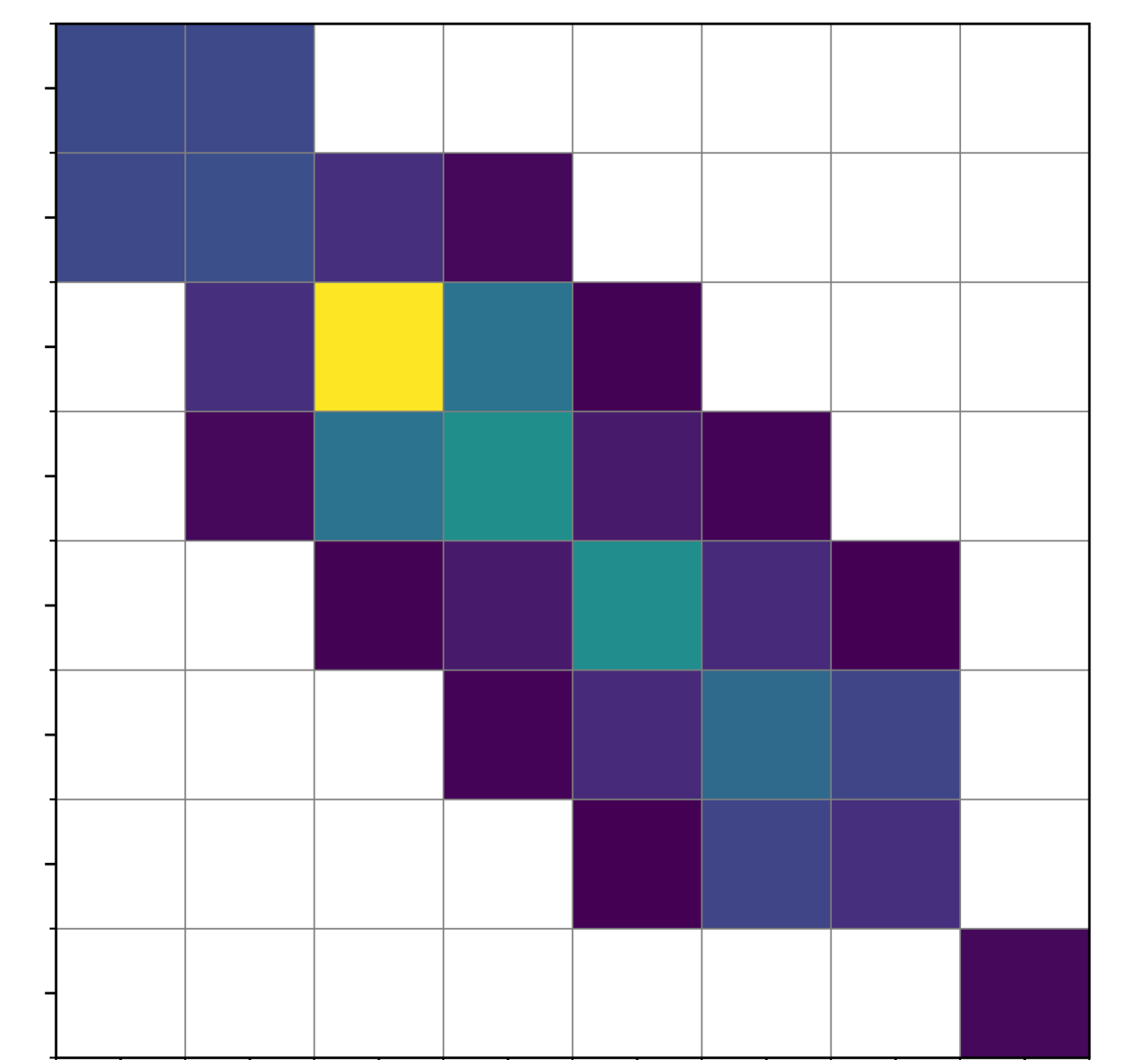
$$\mathbf{K}_{\mathbf{u}\mathbf{u}} = [\langle \phi_i, \phi_j \rangle_{\mathcal{H}}]_{i,j=1}^M$$



$$\mathbf{K}_{\mathbf{u}\mathbf{f}} = [\phi_i(x_j)]_{i=1,j=1}^{M,N}$$



$$\mathbf{K}_{\mathbf{u}\mathbf{f}}\mathbf{K}_{\mathbf{f}\mathbf{u}}$$



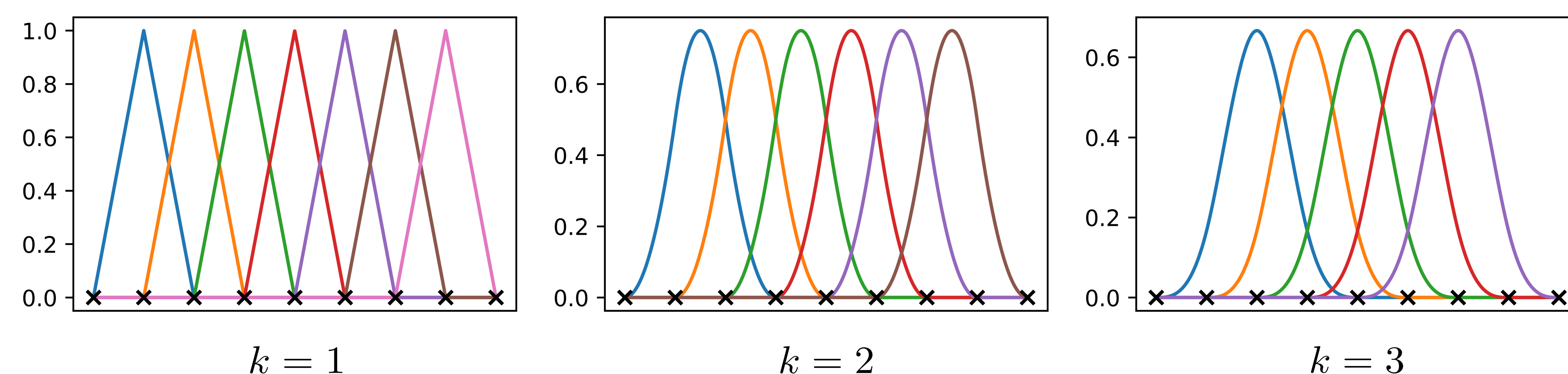
- Projecting onto B-spline basis, $(\mathbf{K}_{\mathbf{u}\mathbf{u}} - \sigma^{-2}\mathbf{K}_{\mathbf{u}\mathbf{f}}\mathbf{K}_{\mathbf{f}\mathbf{u}})$ is a **band diagonal matrix**

B-Spline Inducing Features

- Construct inter-domain inducing variables by projecting GP onto a set of compactly supported basis functions

$$u_m = \langle f, \phi_m \rangle_{\mathcal{H}}$$

- ϕ_m are B-Spline basis functions of order k



Inference

- Cholesky factor of a band diagonal matrix computed in $\mathcal{O}(Mk^2)$ where k is the bandwidth of the matrix

Fit a GP with $N = 2,049,279$ and $M = 1000$ in under 4 seconds!

Method	$M = 1000$	$M = 10,000$	$M = 25,000$
AS-VGP (MSE $\times 10^{-1}$)	8.67	4.53	2.94
SVGP (MSE $\times 10^{-1}$)	8.98	/	/
AS-VGP (NLPD)	1.35	1.04	0.86
SVGP (NLPD)	1.37	/	/
AS-VGP (Time in s)	3.90	18.62	51.69
SVGP (Time in s)	932	/	/