#### Scalable Gaussian Processes<sup>1</sup>

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¹Main papers: Quiñonero Candela and Rasmussen (2005),Hensman et al. (2013) and Wu et al. (2022)

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#### Problem Statement

 Gaussian Processes can accurately model complex data, but they're also very inefficient. Is there a way to make them more scalable?



### Gaussian Processes

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- 0



### Gaussian Processes

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# Sparse GPs

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# Subset of Regressors (SoR)

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- 0



# Sparse Pseudo Input Gaussian Processes (SPGP)

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# Unifying framework for Sparse GP

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### Variational Gaussian Process

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- 0



### Variational Gaussian Process

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- 0



## Stochastic Variational Inference for GPs

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## Stochastic Variational Inference for GPs



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## Variational Nearest Neighbor GPS

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## Final Remarks

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#### References I

- Hensman, J., Fusi, N., and Lawrence, N. (2013). Gaussian processes for big data. *Proceedings of the Twenty-Ninth Conference on Uncertainty in Artificial Intelligence (UAI)*, pages 282–290.
- Quiñonero Candela, J. and Rasmussen, C. E. (2005). A unifying view of sparse approximate gaussian process regression. *Journal of Machine Learning Research*, 6:1939–1959.
- Wu, L., Pleiss, G., and Cunningham, J. P. (2022). Variational nearest neighbor Gaussian process. In *Proceedings of the 39th International Conference on Machine Learning*, volume 162 of *Proceedings of Machine Learning Research*, pages 24114–24130. PMLR.

