# ProbNum: Probabilistic Numerics in Python

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Putting theory into practice.



- Hennig et al. [1] *"Efficient and stable implementations are still in development. Convincing use-cases from various scientific disciplines are only beginning to emerge."*
- Oates and Sullivan [2] "[...] the coming to fruition of this vision will require demonstrable success on problems that were intractable with the computational resources of previous decades [...]."

### Realizing the vision of probabilistic numerics necessitates software.

### Goals

- + demonstrate the value of a probabilistic approach
- + make use of probabilistic numerical methods in application
- + propagate uncertainty through chains of computations



### Software Matters to Research

Software packages advance progress in a research field





Software is a central part of modern science.

### **Key Advantages**

- + Demonstration and prototyping
- Reproducible research
- + Enables new applications

- + Faster method development
- Rapid benchmarking
- + Reveals research questions



### ProbNum

Probabilistic Numerics in Python.





Learn to Approximate. Approximate to Learn.

Get Started 💅



Solve Numerical Problems

Solve problems from linear algebra, optimization, quadrature and differential equations using probabilistic inference.





Quantify and propagate uncertainty from finite resources and stochastic input in computational pipelines.



Compose Custom Numerical Methods

Create problem-specific probabilistic numerical methods from predefined or your own custom components.

### Solving a Linear System

Basic Interface of ProbNum's PN methods.



Ax = b



```
import numpy as np
```

```
x = np.linalg.solve(A, b)
x
```



import probnum as pn

belief = pn.linalg.problinsolve(A, b)
belief.x



## Specifying Prior Information

Encoding prior information from various sources into an algorithm.









belief.from\_spectrum(lmbda)

belief.from\_precond(P)

belief.from\_observations(X)

Belief constructors define common ways of including prior information.

### Composing Custom Methods.

Creating new from old



#### # Construct custom solver

```
pls = pn.linalg.ProbabilisticLinearSolver(
    policy=ConjugateGradientPolicy(),
    information_op=ProjectedRHSInformationOp(),
    belief_update=LinearGaussianInferenceBeliefUpdate(),
    stopping_criterion=PosteriorContractionStopCrit(tol=1e-5),
)
```

```
# Solve problem
pls.solve(problem=(A, b))
```





### Problem: Lotka-Volterra equations (first-order, non-linear ODE)



## What's in it for me?

A community software framework benefits everyone



demo







Source: Teymur et al. [5]

## research & develop



## implement

class ProbabilisticNumericalMethod(ABC, Generic[ProblemType, BeliefType]):
 """Probabilistic numerical methods.

An abstract base class defining the implementation of a probabilistic numerical method  $[1]_{-}$  [2]\_. A PN method solves a numerical problem by treating it as a probabilistic inference task.

#### Parameters

.....

prior :

Prior knowledge about quantities of interest for the numerical problem.

#### References

 [1] Hennig, P., Osborne, Mike A. and Girolami M., Probabilistic numerics and uncertainty in computations. "Proceedings of the Royal Society of London A: Nathematical, Physical and Engineering Sciences, 471(2178), 2015.

... [2] Cockayne, J., Oates, C., Sullivan Tim J. and Girolami M., Bayesian probabilistic numerical methods. "SIAM Review", 61(4):756-789, 2019 apply



Source: Wenger and Hennig [6]

## test & benchmark



### Join the Effort!





- + Solve numerical problems.
- + Quantify uncertainty in computation.
- + Compose custom probabilistic numerical methods.





- Philipp Hennig, Mike A. Osborne, and Mark Girolami. Probabilistic numerics and uncertainty in computations. Proceedings of the Royal Society of London A: Mathematical, Physical and Engineering Sciences, 471(2179), 2015.
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- [3] Michael Schober, Simo Särkkä, and Philipp Hennig. A probabilistic model for the numerical solution of initial value problems. Statistics and Computing, 29(1):99–122, 2019.
- [4] Nathanael Bosch, Philipp Hennig, and Filip Tronarp. Calibrated adaptive probabilistic ODE solvers. In Proceedings of The 24th International Conference on Artificial Intelligence and Statistics (AISTATS), 2021.
- [5] Onur Teymur, Christopher N. Foley, Philip G. Breen, Toni Karvonen, and Chris. J. Oates. Black box probabilistic numerics. arXiv preprint, 2021. URL http://arxiv.org/abs/2106.13718.
- [6] Jonathan Wenger and Philipp Hennig. Probabilistic linear solvers for machine learning. In Advances in Neural Information Processing Systems (NeurIPS), 2020.