
The Bayesian Analysis Toolkit and applications

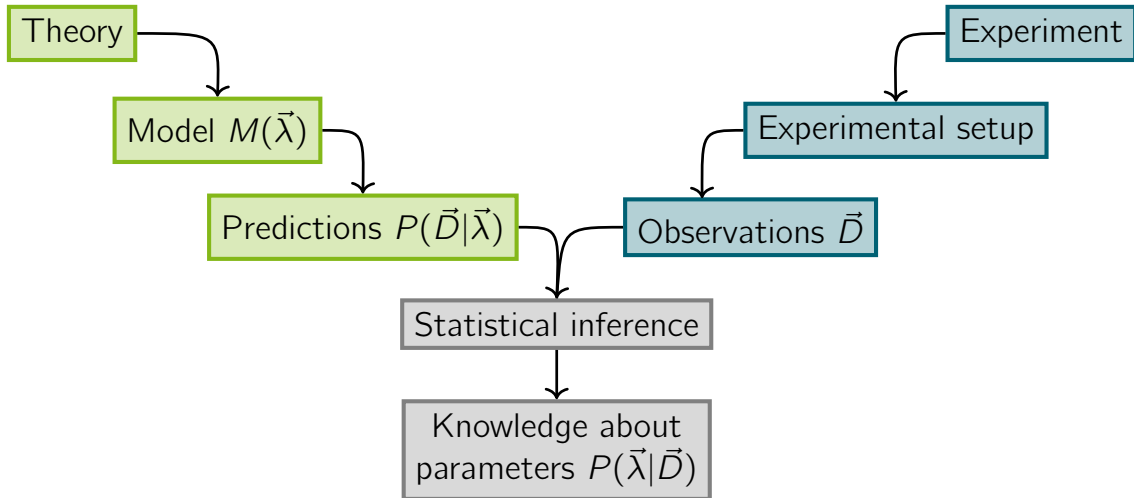
Cornelius Grunwald¹, Kevin Kröninger¹,
Romain Madar², Stéphane Monteil², Lars Röhrig^{1,2}

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¹Department of Physics – TU Dortmund University

²Laboratoire de Physique de Clermont – Université Clermont-Auvergne

What it's all about and who am I?



What is BAT.jl

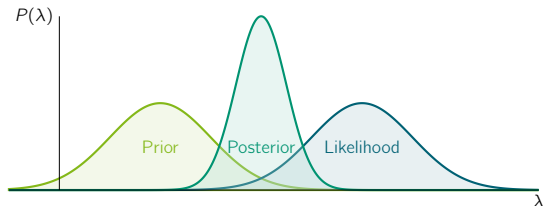
- Toolkit for performing Bayesian inference in a user-friendly way
- Provides a collection of algorithms and methods without relying on specific modeling language
- Focusing on sampling custom posterior distributions
- Further functionalities:
 - Integration & marginalisation
 - Optimisation & parameter estimation
 - Limit setting, model comparison, goodness-of-fit tests

Bayes' Theorem

$$P(\lambda|D) = \frac{P(D|\lambda) \cdot P(\lambda)}{\int P(D|\lambda) \cdot P(\lambda) d\lambda}$$

D : data

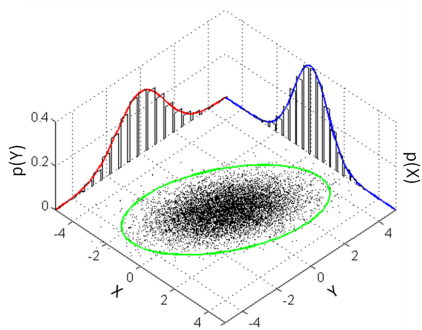
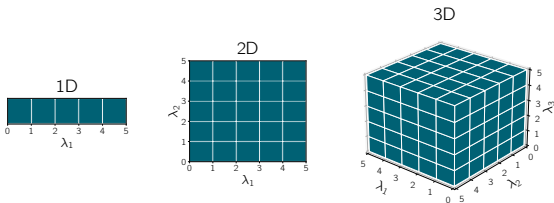
λ : parameters



The need for numerical methods

The curse of dimensionality:

- In high-dimensional parameter spaces: unfeasible to evaluate the posterior at all points of the sampling-space

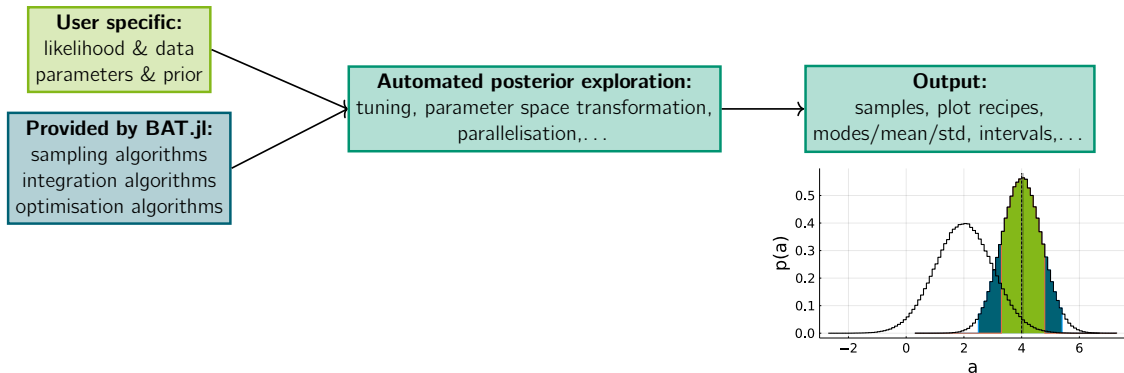


- Efficient algorithms for high-dimensional

- ... sampling
- ... optimisation
- ... integration

BAT.jl – The Bayesian Analysis Toolkit

- Originally developed in C++, but not maintained anymore
- BAT.jl: Rewrite in Julia programming language, first release in 2019
- Modern software package with toolkit-like character for easy expansion



The Julia language

- Language designed for high-performance and efficient numerical computing
- First launch 2012 after 3 years of development at MIT
- v1.0 in 2018, current v1.7.3
- Solved the two-language-problem:
 - As comfortable as python
 - As fast as C++
- Key features: dynamic type system, multiple dispatch, parallel & distributed computing, package manager, easy to call Fortran, C/C++, python,...
- Growing and very scientific community

```
(base) [~] $ julia
┌───────────┴───────────┐
│  Documentation: https://docs.julialang.org
│  Type "?" for help, "]"? for Pkg help.
│  Version 1.7.2 (2022-02-06)
│  Official https://julialang.org/ release
└───────────┬───────────┘
julia> function add(x, y)
           return x + y
        end
add (generic function with 1 method)

julia> s = add(1, 2)
3
```

Learn julia [here](https://docs.julialang.org).

Features of BAT.jl

- Use of custom posterior distributions (from user-specific likelihoods & priors)
- Collection of sampling algorithms:
 - MCMC: Metropolis-Hastings, Hamiltonian-MC
 - Importance Samplers
 - Nested Sampling
- Automated initialisation, tuning & convergence testing for MC chains
- Automated parameter space transformations
- Design idea: offer reasonable default settings for easy-to-use, but also allow for fine-grained control

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│  └───┘ └───┘ └───┘    │
└───────────┴───────────┘
julia> using Pkg
julia> Pkg.add("BAT")
```

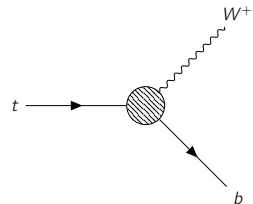
The package is available at GitHub [here](#).

Let's explore together!

You're warmly invited to try yourself with **this** binder!

For what BAT.jl is used (among others): EFTfitter.jl

- Tool for combining multiple measurements & performing Bayesian inference on the underlying parameters
- Optimised for EFT interpretations of measurements with interface to BAT.jl
- SMEFT: higher-dimensional operators extending the SM-Lagrangian, can affect e.g. t -quark decay



Effective coupling at $t \rightarrow Wb$.

SMEFT-likelihood

$$\ln(\mathcal{L}(\vec{x} | \vec{y}(\vec{\lambda}))) = \sum_{i,j=1}^{\overbrace{n}^{\text{\# measurements}}} \underbrace{[\vec{x} - \vec{y}(\vec{\lambda})]_i}_{\text{distance between measured \& predicted values}} \overbrace{\mathcal{M}_{ij}^{-1}}^{\text{covariance matrix}} [\vec{x} - \vec{y}(\vec{\lambda})]_j$$

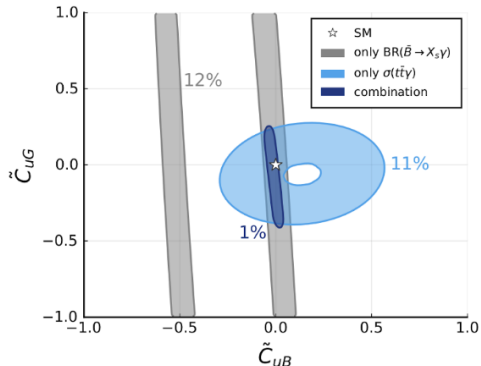
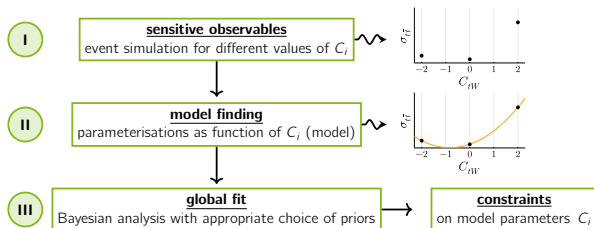
x : measured σ

λ : Wilson coefficients

y : pred. $\sigma(\lambda)$

For what BAT.jl is used (among others): EFTfitter.jl

- Simulate predicted cross-sections with the Monte-Carlo generator MadGraph
- Parameterise the **simulations** according to $y(\lambda)$
- Compare with **measurements** from HEP collaborations, e.g. ATLAS, BELLE, ...
- Choosing flat priors on the **model parameters** to constrain, e.g. C_{UG} , C_{UB} and C_{UW}



Fit of t - and b -physics observables [1].

Conclusions and how to get started...

- BAT software highly efficient in `julia` programming language and provides a variety of algorithms for sampling, optimisation and integration
- Known weaknesses of existing sampling algorithms + modularity = room for contributions

- Package as GitHub repository [available](#)
- Full (API) [documentation](#), as well as [tutorials](#)
- Implementation of new sampling algorithms in modular-like fashion, see e.g. [different importance sampler](#)

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Thanks a lot for your attention!