

A dark blue background featuring a complex network of glowing, multi-colored lines representing particle trajectories or light paths. Several bright, glowing points of varying colors (blue, green, yellow, orange) are scattered throughout the scene, some with lens flare effects.

zfit

scalable pythonic fitting

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FNSNF

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What is zfit?

zfit
scalable pythonic fitting

Fitting for HEP
(~*RooFit Core*)

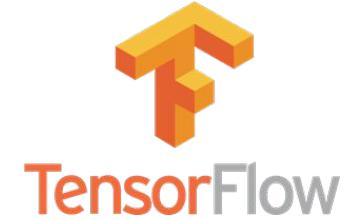
Fresh API, codebase
(*inspired by others*)

scalable pythonic fitting

Scalable

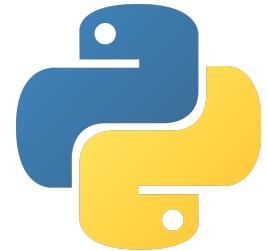
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scalable pythonic fitting

- TensorFlow **hidden** backend, uses graphs
 - numpy-like syntax
 - parallelization on CPU/GPU, analytic gradient,...
- Writing functions simple for users *and* developers
 - No Cython, MPI, CUDA,... for *state-of-the-art performance*
 - No low-level maintenance required!
- Used in multiple physics libraries and analyses



Pythonic

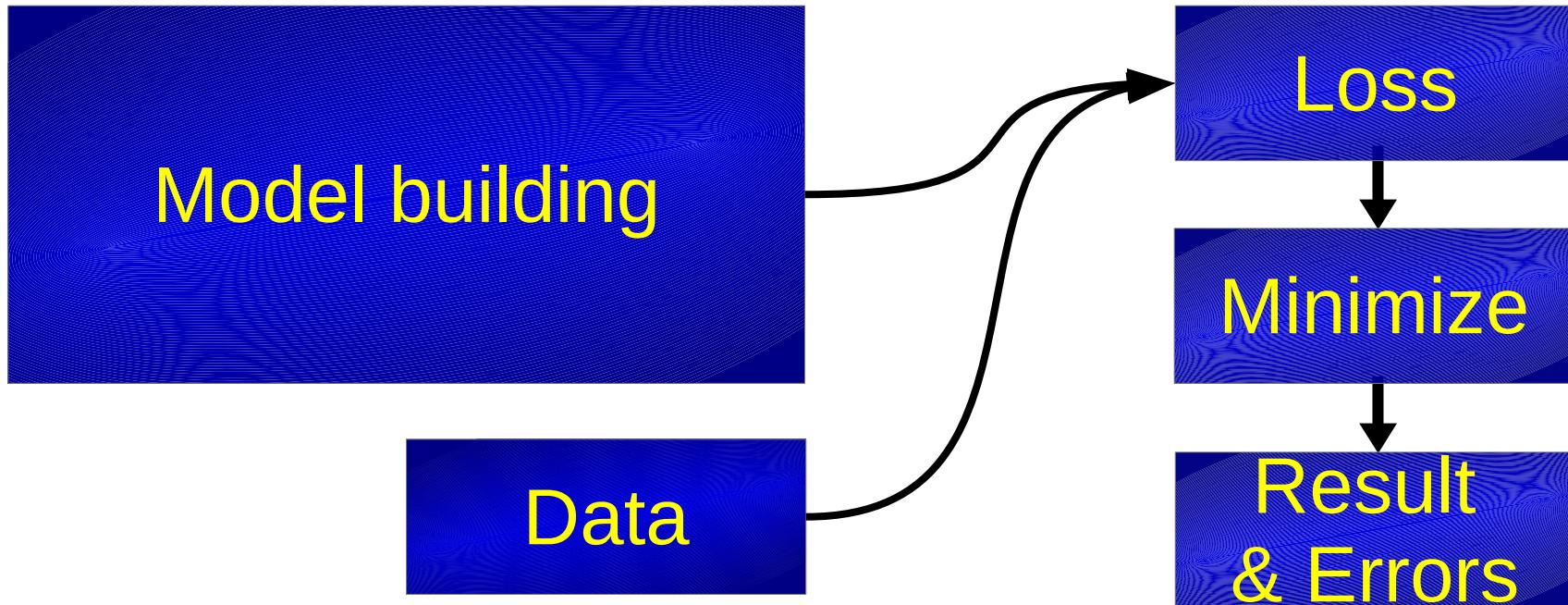
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- Pure Python
- Integrated into python ecosystem
 - Focus *only* on one thing
 - Integrate seamlessly with data, plotting and statistic packages
 - Uses [uproot](#) for ROOT files (no ROOT dependency!)
 - Used by [lauztat](#) for statistics
 - Can be used by [phasespace](#) for resonance modeling
- Easily extendable
 - e.g. custom PDF

Fitting

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Complete fit

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```
import zfit

normal_np = np.random.normal(loc=2., scale=3., size=10000)

obs = zfit.Space("x", limits=(-10, 10))

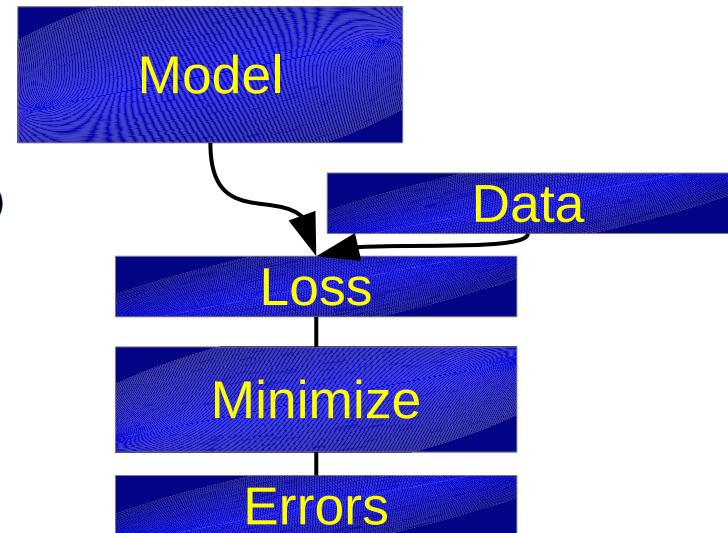
mu    = zfit.Parameter("mu", 1., -5, 5)
sigma = zfit.Parameter("sigma", 3., 1, 10)
gauss = zfit.pdf.Gauss(mu=mu, sigma=sigma, obs=obs)

data = zfit.data.Data.from_numpy(obs=obs, array=normal_np)

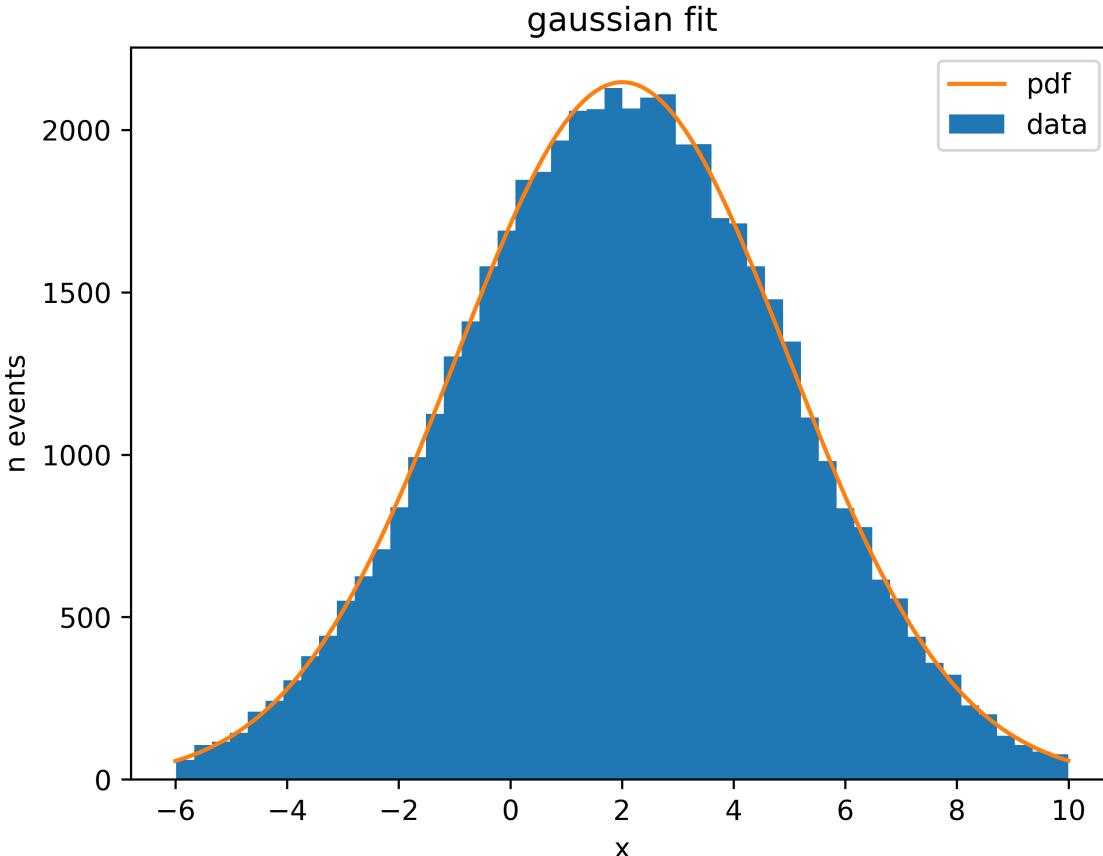
nll = zfit.loss.UnbinnedNLL(model=gauss, data=data)

minimizer = zfit.minimize.MinuitMinimizer()
result = minimizer.minimize(nll)

param_errors = result.error()
```



Complete fit: plot



Gaussian PDF fitted to
gaussian distributed sample

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Model building

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```
obs = zfit.Space("x", limits=(-10, 10))

mu = zfit.Parameter("mu", 1, -4, 6)
sigma = zfit.Parameter("sigma", 1, 0.1, 10)
lambd = zfit.Parameter("lambda", -1, -5, 0)
frac = zfit.Parameter("fraction", 0.5, 0, 1) }
```

parameters

```
gauss = zfit.pdf.Gauss(mu=mu, sigma=sigma, obs=obs)
exponential = zfit.pdf.Exponential(lambd, obs=obs) }
```

models

```
sum_pdf = zfit.pdf.SumPDF([gauss, exponential], fracs=frac)
```

```
sum_pdf = frac * gauss + exponential
```

equivalent

Custom PDF

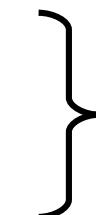
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```
from zfit import ztf
```

```
class CustomPDF(zfit.pdf.ZPDF):
    _PARAMS = ['alpha']

    def _unnormalized_pdf(self, x):
        data = x.unstack_x()
        alpha = self.params['alpha']

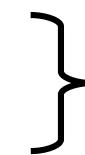
        return ztf.exp(alpha * data)
```



implement custom function

```
custom_pdf = CustomPDF(obs=obs, alpha=0.2)
```

```
integral = custom_pdf.integrate(limits=(-1, 2))
sample   = custom_pdf.sample(n=1000)
prob     = custom_pdf.pdf(sample)
```



use functionality of model

Simultaneous fit

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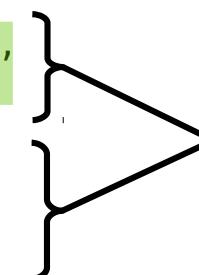
```
mu_shared = zfit.Parameter("mu_shared", 1., -4, 6)
sigma1 = zfit.Parameter("sigma_one", 1., 0.1, 10)
sigma2 = zfit.Parameter("sigma_two", 1., 0.1, 10)

gauss1 = zfit.pdf.Gauss(mu=mu_shared, sigma=sigma1, obs=obs)
gauss2 = zfit.pdf.Gauss(mu=mu_shared, sigma=sigma2, obs=obs)
```

} shared parameters

```
nll_simultaneous = zfit.loss.UnbinnedNLL(model=[gauss1, gauss2],
                                             data=[data1, data2])
```

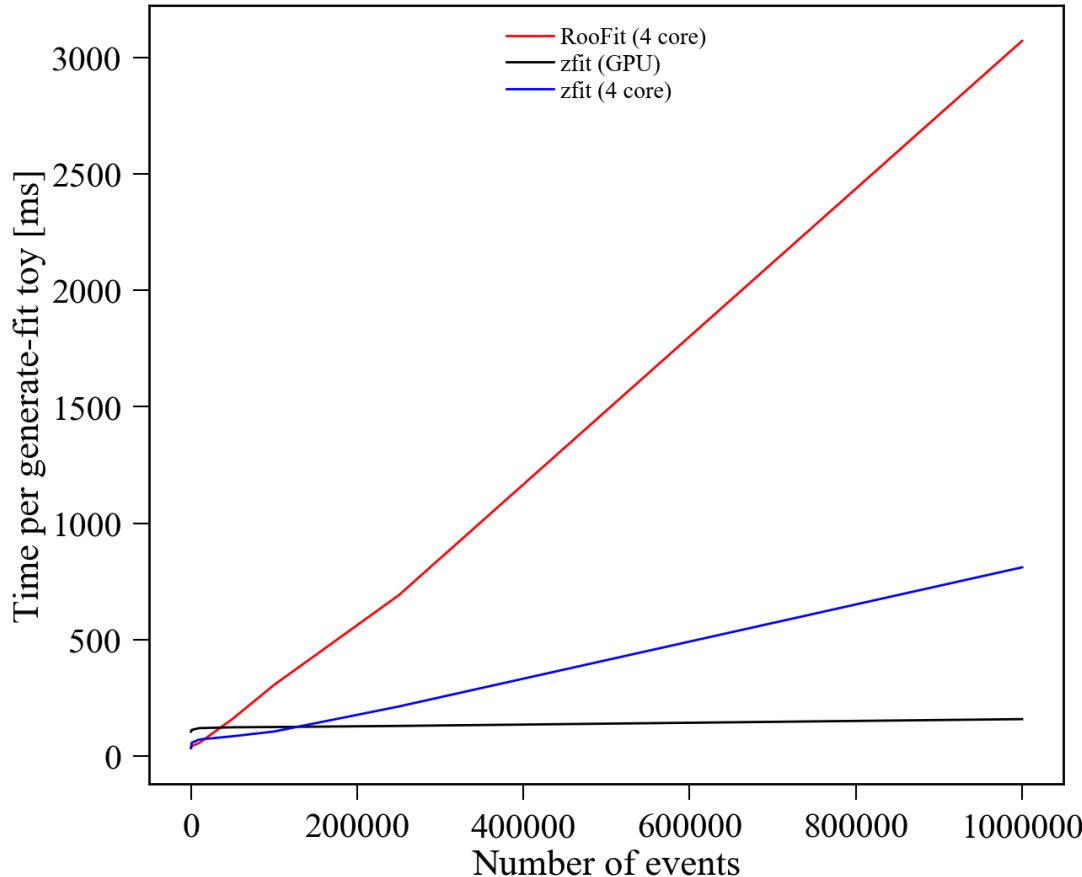
```
nll1 = zfit.loss.UnbinnedNLL(model=gauss1, data=data1)
nll2 = zfit.loss.UnbinnedNLL(model=gauss2, data=data2)
nll_simultaneous2 = nll1 + nll2
```



Completely
equivalent

Simple toy fit: speed

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Multiple sample and fit

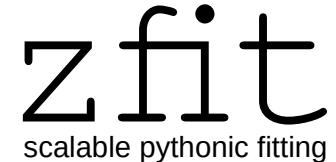
Comparison

RooFit (4 cores)

zfit (4 cores)

zfit (GPU)

Summary



- Beta stage, **usable!** (already used in LHCb analyses)
 - Not feature complete, but API stabilizing
- Contributions in form of *feedback and criticism* very welcome
 - API, use-cases, bugs,...
 - Any crazy idea!

It's about a reliable library
not about
«we need to get it working fast»

<https://zfit.github.io/zfit/>

zfit@GitHub



Gitter channel



zfit@physik.uzh.ch

Join the discussion!

scalable

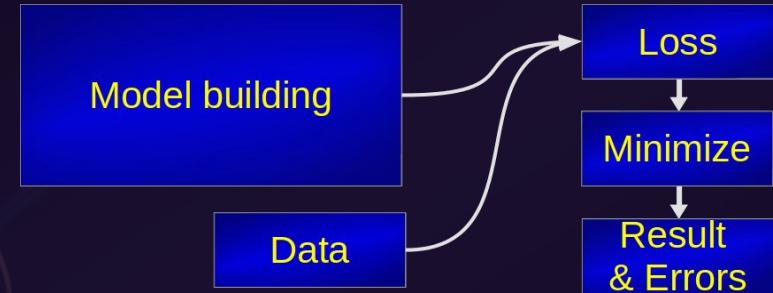


TensorFlow

pythonic



fitting



zfit

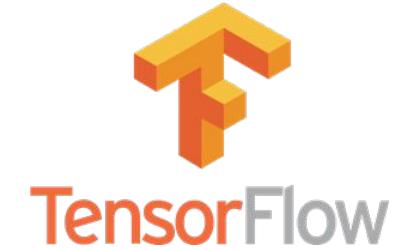
It's about a reliable library

Backup Slides

Scalable: TensorFlow

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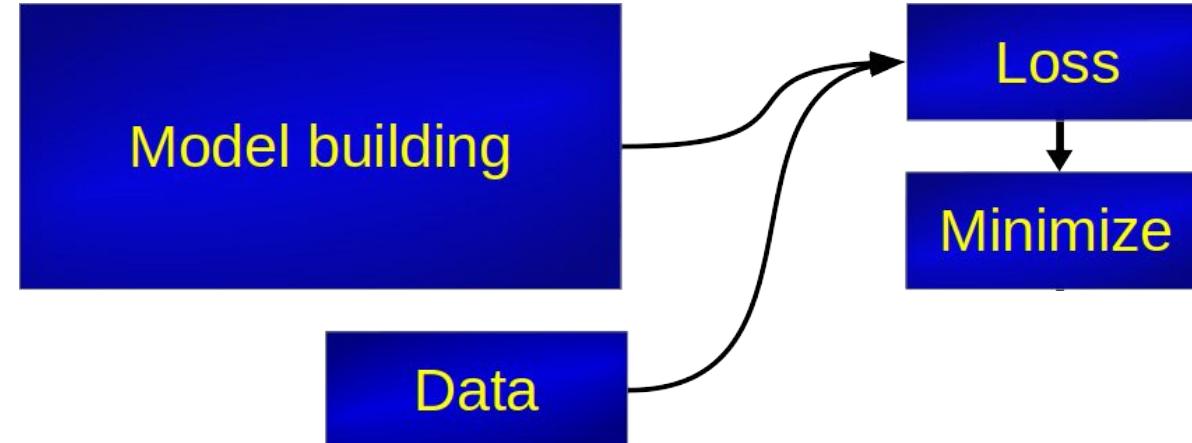
- Deep Learning framework by Google
- Modern, declarative graph approach
- Built for highly parallelized, fast communicating CPU, GPU, TPU,... clusters
- Built to use «Big Data»



Scalable: TensorFlow

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- Machine learning in a nutshell:
 - Build a model (a lot of matrix multiplications with simple non-linear functions in between) with 100k+ free parameters
 - Create a loss function (see how good/bad the predictions are)
 - Minimize it





Pythonic: statistics tool «lauztat»

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- Author: Matthieu Marinangeli
- WIP, pre-beta
- Python statistics tool for limits, significance etc.
(~ RooStats)
- [lauztat on Github](#) with [example notebooks](#) using zfit



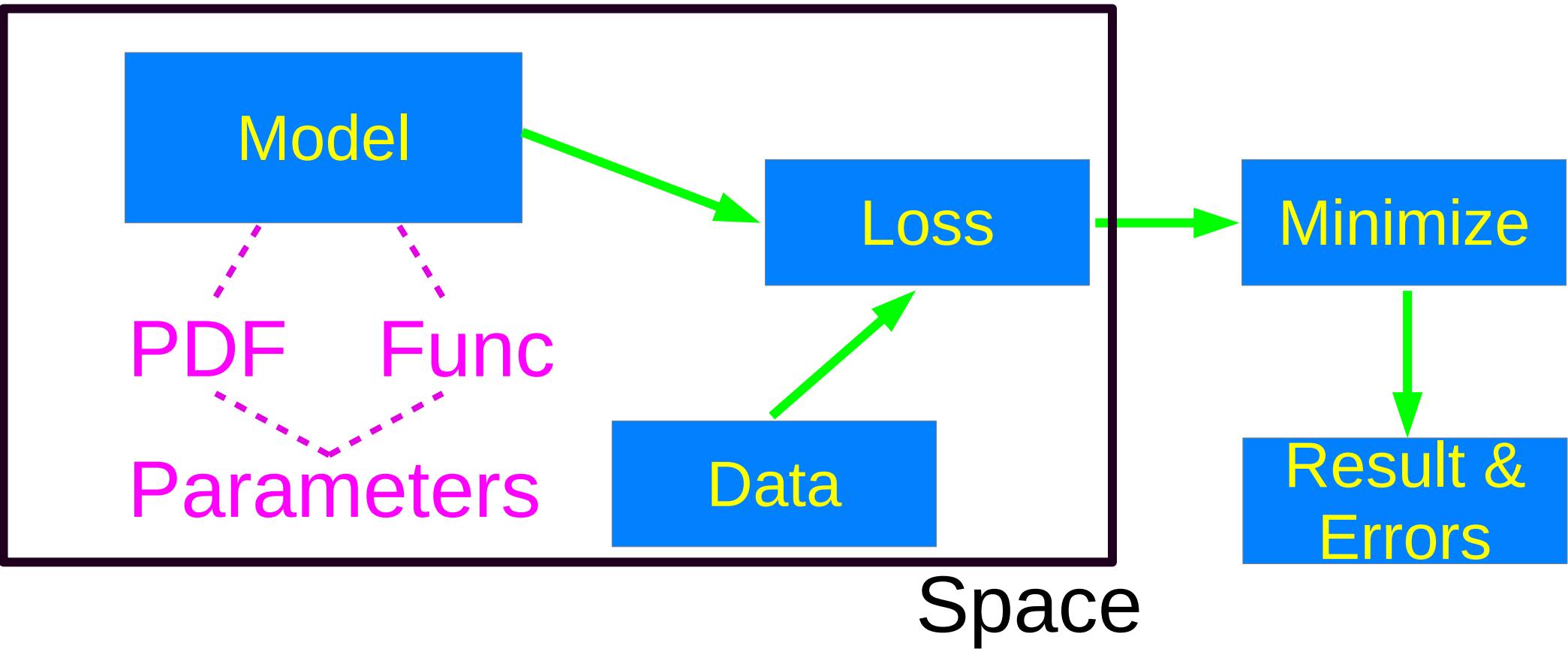
Pythonic: «phasespace»

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- Author: Albert Puig
- Python tool for n-body phasespace generation (~ TgenPhaseSpace)
- Uses TensorFlow as backend
- Can use zfit pdf for sampling

Fitting: complete structure

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Performance toy example

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