

# PyMC library

open source Probabilistic Programming framework

# Sum Up

PyMC (formerly known as PyMC3) is a Python package for Bayesian statistical modeling and probabilistic machine learning which focuses on advanced Markov chain Monte Carlo and variational fitting algorithms.

Coding Bayesian statistical models the statistician way.

## Installation / Import

- `conda create -c conda-forge -n pymc_env "pymc>=4"`
- `conda activate pymc_env`

```
>> import pymc as pm
```

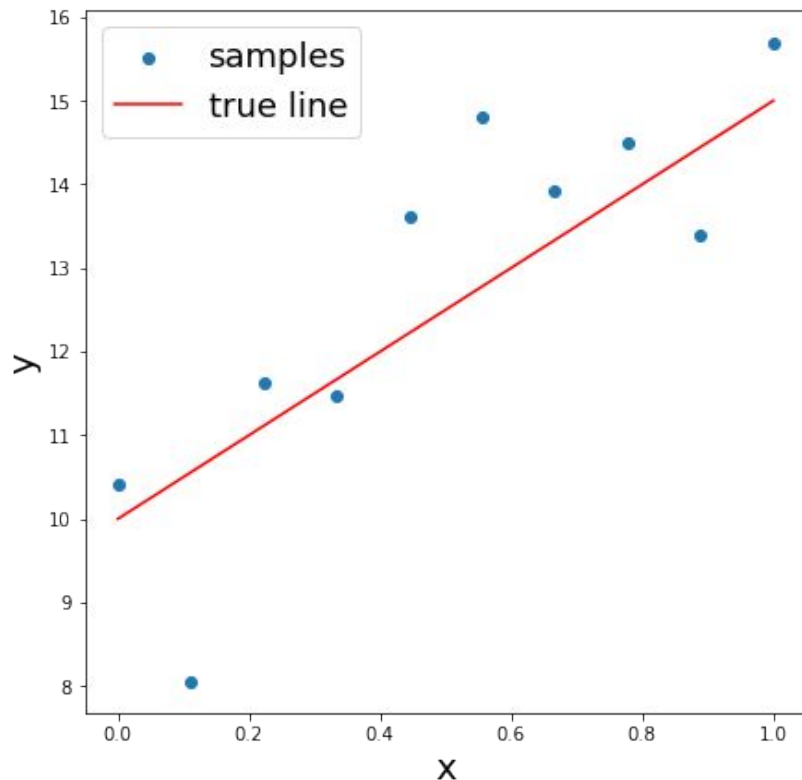
# Usual way on a small dataset (10 points)

**Case of linear regression:**

$$y_{\text{true}} = mx + b$$

**Fitting an usual linear model:**

$$y_{\text{obs}} = y_{\text{true}} + N(0, \sigma)$$



# Sample distribution way

Let's consider the following independent priors:

- $m \sim N(0,20)$
- $b \sim N(0,20)$
- $\sigma \sim \text{Exp}(1)$

Likelihood :  $y \mid m, b, \sigma \sim N(mx+b, \sigma)$

Posterior :  $m, b, \sigma \mid y \sim ?$

Bayesian theorem gives us:

$$P(m, b, \sigma \mid y) \propto P(y \mid m, b, \sigma) \times P(m) \times P(b) \times P(\sigma)$$

PyMC will use priors and likelihood to compute a sample of the Posterior

```
with pm.Model() as model:
    #priors
    sigma = pm.Exponential("sigma", lam=1.0)
    intercept = pm.Normal("intercept", mu=0, sigma=20)
    slope = pm.Normal("slope", mu=0, sigma=20)

    #Likelihood
    likelihood = pm.Normal("y", mu=slope*x_vals + intercept, sigma=sigma, observed=y)

    #posterior
    trace = pm.sample(1000, cores=4)
```

# Result

