

Introduction to Monte Python

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Outline

- 1 Monte Python
 - Goals
 - Design Strategy

- 2 Basic Usage
 - Installation
 - Usage
 - Running strategies

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- 2 Basic Usage

Goals of Monte Python

Wish list

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Guidelines

- **Modular structure**: separate I/O, parser, data structures, sampler, interface with the cosmological module, plotting: **Tuesday**

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- **Convenient Plotting**: since a folder will be self contained, with all the information, producing a plot out of this folder should be easy.
- **Using mock data**: there must be a way to handle mock data easily.

Conclusion on design

You tell me !

We'll see in the coming days if it works !

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Installation

Compile the wrapper

```
cd class; make;  
cd python; python setup.py install --user
```

Download

```
http://montepython.net or  
https://github.com/audren/montepython\_public/releases
```

Unzip

```
bunzip2 montepython_v2.0.2.tar.bz2  
tar -xvf montepython_v2.0.2.tar
```

Configure

```
cp default.conf.template default.conf  
edit it to match your path
```

Parser

Common arguments

```
python montepython/MontePython.py run plus...
```

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- `-N 1000` **Number of proposed steps**

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Common arguments

```
python montepython/MontePython.py run plus...
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- *A minima*: `-o chains/planck -p example.param`
- `-N 1000`
- `-c covmat/old.covmat` **better proposal density**

Parser

Common arguments

```
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```

- *A minima*: `-o chains/planck -p example.param`
- `-N 1000`
- `-c covmat/old.covmat`
- `-j fast` **for a fast Cholesky decomposition sampling**

Parser

Common arguments

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- *A minima*: `-o chains/planck -p example.param`
- `-N 1000`
- `-c covmat/old.covmat`
- `-j fast` for a fast Cholesky decomposition sampling
- `-m NS` to use **MultiNest and the Nested Sampling algorithm**

Input Parameter File

```
data.experiments=['fake_planck_bluebook']

# Cosmological parameters list
data.parameters['omega_b'] = [2.249, None, None, 0.016, 0.01, 'cosmo']
data.parameters['omega_cdm'] = [0.1120, None, None, 0.0016, 1, 'cosmo']
data.parameters['n_s'] = [0.963, None, None, 0.004, 1, 'cosmo']
data.parameters['A_s'] = [2.42, None, None, 0.038, 1e-9, 'cosmo']
data.parameters['h'] = [0.703, None, None, 0.0065, 1, 'cosmo']
data.parameters['tau_reio'] = [0.085, None, None, 0.0044, 1, 'cosmo']

# Derived parameter list
data.parameters['z_reio'] = [0, -1, -1, 0, 1, 'derived']
data.parameters['Omega_Lambda'] = [0, -1, -1, 0, 1, 'derived']

data.cosmo_arguments['N_eff'] = 3.046

data.N=10

data.write_step=5
```

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Mean values

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Lower-Upper bounds

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1σ proposal

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scale

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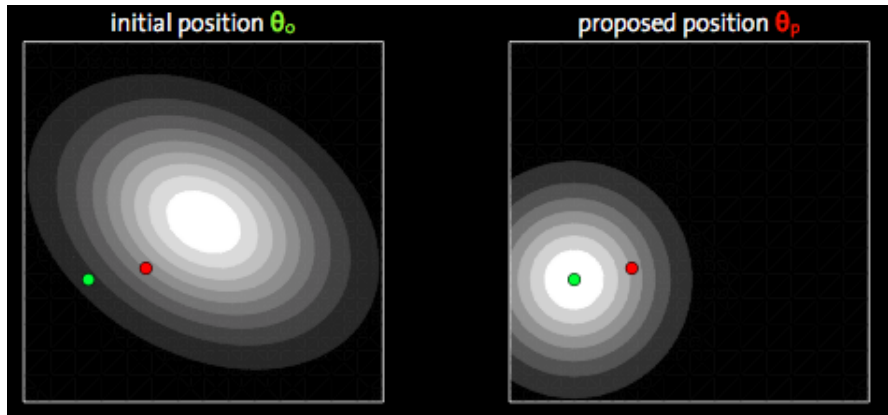
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Number of steps, write buffer

Reminder from lecture 1



Types of parameter

cosmological

- **known to Class (can define tricks)**
- **must be compatible (as in `explanatory.ini`)**

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nuisance

- **As the name indicates...**
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nuisance

- As the name indicates. . .
- **Needed pain**

Warning!

Input parameter file and log.param

- When a folder is **created**, a file `log.param` is written, copying information from param and likelihoods.
- When a new chain is launched, **the input file is not read any more**, only the `log.param`

Configuration File

```
root = '/Users/benjaminAudren/Desktop/professional/codes'  
path['cosmo'] = root+'/class/'
```

Usage: summary

After installation

- `cp default.conf.template default.conf` and edit
- `python montepython/MontePython.py -o chains/test -p example.param`

Running strategies

Starting

- Choosing **experiments** to combine - **careful with that**
- Varying **parameters**, proposal distribution
- After M ($\simeq 10$) chains of N ($\simeq 10000$) points, **analyze**

Analyzing

```
python montepython/MontePython.py info folder
```

Main

- Feed the new found **covariance matrix** to new runs.
- Analyze **only these new chains**
- Compare best-fit likelihood, or evidence, read parameter constraints

What Monte Python does for you

Convenience

- No need to choose a name for your chain

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- Can analyze only certain chains