

# pyobs Documentation Release 1.0.0

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## THE OBS CLASS

class pyobs.obs(orig=None, desc='unknown') Class defining an observable

#### **Parameters**

- orig (obs, optional) creates a copy of orig
- desc (str, optional) description of the observable

#### **Examples**

```
>>> from pyobs import obs
>>> a = obs(desc='test')
```

add\_syst\_err(name, err)

Add a systematic error to the observable

## **Parameters**

- name (str) label that uniquely identifies the syst. error
- err (array) array with the same dimensions of the observable with the systematic error

## **Examples**

 $\label{lem:create} \begin{tabular}{ll} $create(ename,\ data,\ icnfg=None,\ rname=None,\ shape=1,\ lat=None) \\ Create an observable \\ \end{tabular}$ 

## Parameters

- ename (str) label of the ensemble
- $\bullet$  data (array , list of arrays ) the data generated from a single or multiple replica
- ullet icnfg (array of ints or list of arrays of ints, optional)—indices of the configurations corresponding to data; if not passed the measurements are assumed to be contiguous

- rname (str or list of str, optional) identifier of the replica; if not passed integers from 0 are automatically assigned
- shape (tuple, optional) shape of the observable, data must be passed accordingly
- lat (list of ints, optional) the size of each dimension of the master-field; if passed data is assumed to be obtained from observables measured at different sites and icnfg is re-interpreted as the index labeling the sites; if icnfg is not passed data is assumed to be contiguous on all sites.

**Note:** For data and icnfg array can mean either a list or a 1-D numpy.array. If the observable has already been created, calling create again will add a new replica to the same ensemble.

#### **Examples**

```
>>> data = [0.43, 0.42, ...] # a scalar observable
>>> a = pyobs.obs(desc='test')
>>> a.create('EnsembleA',data)
```

```
>>> data0 = [0.43,0.42, ...] # replica 0

>>> data1 = [0.40,0.41, ...] # replica 1

>>> a = pyobs.obs(desc='test')

>>> a.create('EnsembleA',[data0,data1],rname=['r0','r1'])
```

```
>>> data = [0.43, 0.42, 0.44, ...]
>>> icnfg= [ 10, 11, 13, ...]
>>> a = pyobs.obs(desc='test')
>>> a.create('EnsembleA',data,icnfg=icnfg)
```

```
>>> data = [1.0, 2.0, 3.0, 4.0, ...]
>>> a = pyobs.obs(desc='matrix')
>>> a.create('EnsembleA',data,shape=(2,2))
```

## **Examples**

```
>>> data = [0.43, 0.42, 0.44, ...]
>>> lat = [64,32,32,32]
>>> a = pyobs.obs(desc='test-mf')
>>> a.create('EnsembleA',data,lat=lat)
```

```
>>> data = [0.43, 0.42, 0.44, ...]
>>> idx = [0, 2, 4, 6, ...] # measurements on all even points of time-slice
>>> lat = [32, 32, 32]
>>> a = pyobs.obs(desc='test-mf')
>>> a.create('EnsembleA',data,lat=lat,icnfg=idx)
```

create\_from\_cov(cname, value, covariance)

Create observables based on covariance matrices

#### Parameters

- cname (str) label that uniquely identifies the data set
- value (array) central value of the observable; only 1-D arrays are supported
- covariance (array) a 2-D covariance matrix; if covariance is a 1-D array of same length as value, a diagonal covariance matrix is assumed.

## **Examples**

```
>>> mpi = pyobs.obs(desc='pion masses, charged and neutral')
>>> mpi.create_cd('mpi-pdg18',[139.57061,134.9770],[0.00023**2,0.0005**2])
>>> print(mpi)
139.57061(23) 134.97700(50)
```

```
error(errinfo=\{\}, plot=False, pfile=None)
```

Estimate the error of the observable, by summing in quadrature the systematic errors with the statistical errors computed from all ensembles and master fields.

#### **Parameters**

- errinfo (dict, optional) dictionary contains one instance of the errinfo class for each ensemble/master-field. The errinfo class provides additional details for the automatic or manual windowing procedure in the Gamma method. If not passed default parameters are assumed.
- plot (bool, optional) if specified a plot is produced, for every element of the observable, and for every ensemble/master-field where the corresponding element has fluctuations. In addition one piechart plot is produced for every element, showing the contributions to the error from the various sources, only if there are multiple sources, ie several ensembles. It is recommended to use the plotting function only for observables with small dimensions.
- pfile (str, optional) if specified all plots produced with the flag plot are saved to disk, using pfile as base name with an additional suffix.

**Returns** the central value and error of the observable.

Return type list of two arrays

Note: By default, the errors are computed with the Gamma method, with the Stau parameter equal to 1.5. Additionally the jackknife method can be used by passing the appropriate errinfo dictionary with argument bs set to a non-zero integer value. For master fields the error is computed using the master-field approach and the automatic windowing procedure requires the additional argument k (see main documentation), which by default is zero, but can be specified via the errinfo dictionary. Through the errinfo dictionary the user can treat every ensemble differently, as explained in the examples below.

#### **Examples**

```
>>> obsA = pyobs.obs('obsA')
>>> obsA.create('A',dataA) # create the observable A from ensemble A
>>> [v,e] = obsA.error() # collect central value and error in v,e
>>> einfo = {'A': errinfo(Stau=3.0)} # specify non-standard Stau for ensemble A
>>> [_,e1] = obsA.error(errinfo=einfo)
>>> print(e,e1) # check difference in error estimation
```

```
>>> obsB = pyobs.obs('obsB')
>>> obsB.create('B',dataB) # create the observable B from ensemble B
>>> obsC = obsA * obsB # derived observable with fluctuations from ensembles A,B
>>> einfo = {'A': errinfo(Stau=3.0), 'B': errinfo(W=30)}
>>> [v,e] = obsC.error(errinfo=einfo,plot=True)
```

```
error_of_error(errinfo={})
```

Returns the error of the error based on the analytic prediction obtained by U. Wolff.

Parameters errinfo (dict, optional) - see the documentation of the error method.

**Returns** the error of the error

Return type array

## peek()

Display a brief summary of the content of the observable, including its memory footprint and requirements (for error computation), its description and ensemble/replica content

## Example

```
>>> obs.peek()
Observable with shape = (1, 4)
- description: vector-test
- size: 82 KB
- mean: [[0.20007161 0.40085252 1.19902686 1.60184989]]
- Ensemble A
- Replica 0 with mask [0, 1, 2, 3] and ncnfg 500
temporary additional memory required 0.015 MB
```

## tauint()

Estimates the integrated autocorrelation time and its error for every ensemble, with the automatic windowing procedure.

## MANIPULATION OF OBSERVABLES

$reshape(x, new\_shape)$	Change the shape of the observable
concatenate(x, y[, axis])	Join two arrays along an existing axis
$transpose(\mathbf{x}[,\mathrm{axes}])$	Transpose a tensor along specific axes.
$sort(\mathbf{x}[, axis])$	Sort a tensor along a specific axis.
diag(x)	Extract the diagonal of 2-D array or construct a
	diagonal matrix from a 1-D array

# 2.1 pyobs.reshape

pyobs.reshape(x, new\_shape)
Change the shape of the observable

#### **Parameters**

- $\bullet$  x (obs) observables to be reshaped
- ullet new\_shape (tuple) the new shape of the observable

 ${\bf Returns} \ {\bf reshaped \ observable}$ 

Return type obs

## Notes

This function acts exclusively on the mean value.

# 2.2 pyobs.concatenate

pyobs.concatenate(x, y, axis=0)

Join two arrays along an existing axis

## Parameters

- y (x,) the two observable to concatenate
- ullet axis (int, optional) the axis along which the observables will be joined. Default is 0.

**Returns** the concatenated observable

#### **Notes**

If x and y contain information from separate ensembles, they are merged accordingly by keeping only the minimal amount of data in memory.

## 2.3 pyobs.transpose

pyobs.transpose(x, axes=None)

Transpose a tensor along specific axes. For an array a with two axes, gives the matrix transpose.

## Parameters

- x (obs) input observable
- axes (tuple or list of ints, optional) If specified, it must be a tuple or list which contains a permutation of [0,1,...,N-1] where N is the number of axes of x. For more details read the documentation of numpy.transpose

**Returns** the transposed observable

Return type obs

## 2.4 pyobs.sort

```
pyobs.sort(x, axis=- 1)
```

Sort a tensor along a specific axis.

#### Parameters

- x (obs) input observable
- axis (int, optional) the axis which is sorted. Default is -1, the
- axis. (last) -

Returns the sorted observable

Return type obs

# 2.5 pyobs.diag

```
pyobs.diag(x)
```

Extract the diagonal of 2-D array or construct a diagonal matrix from a 1-D array

Parameters x (obs) – input observable

Returns the diagonally projected or extended observable

## **UNARY OPERATIONS**

## 3.1 Linear functions

sum(x[, axis])	Sum of array elements over a given axis.
trace(x[, offset, axis1, axis2])	Return the sum along diagonals of the array.

## 3.1.1 pyobs.sum

```
pyobs.sum(x, axis=None)
Sum of array elements over a given axis.
```

#### **Parameters**

- x (obs) array with elements to sum
- axis (None or int or tuple of ints, optional) Axis or axes along which a sum is performed. The default, axis=None, will sum all elements of the input array.

Returns sum along the axis

Return type obs

## **Examples**

```
>>> import pyobs
>>> pyobs.sum(a)
>>> pyobs.sum(a,axis=0)
```

## 3.1.2 pyobs.trace

```
pyobs.trace(x, offset=0, axis1=0, axis2=1)
Return the sum along diagonals of the array.
```

## **Parameters**

- x (obs) observable whose diagonal elements are taken
- offset (int, optional) offset of the diagonal from the main diagonal; can be both positive and negative. Defaults to 0.
- axis2 (axis1,) axes to be used as the first and second axis of the 2-D subarrays whose diagonals are taken; defaults are the first two axes of x.

**Returns** the sum of the diagonal elements

#### **Notes**

If x is 2-D, the sum along its diagonal with the given offset is returned, i.e., the sum of elements x[i,i+offset] for all i. If x has more than two dimensions, then the axes specified by axis1 and axis2 are used to determine the 2-D sub-arrays whose traces are returned. The shape of the resulting array is the same as that of a with axis1 and axis2 removed.

## **Examples**

```
>>> tr = pyobs.trace(mat)
```

# 3.2 Exponential functions

$log(\mathbf{x})$	Return the Natural logarithm element-wise.
exp(x)	Return the exponential element-wise.
cosh(x)	Return the Hyperbolic cosine element-wise.
$sinh(\mathbf{x})$	Return the Hyperbolic sine element-wise.
arccosh(x)	Return the inverse Hyperbolic cosine element-
	wise.

## 3.2.1 pyobs.log

pyobs.log(x)

Return the Natural logarithm element-wise.

Parameters x (obs) – input observable

**Returns** the logarithm of the input observable, element-wise.

Return type obs

## **Examples**

```
>>> logA = pyobs.log(obsA)
```

## 3.2.2 pyobs.exp

pyobs.exp(x)

Return the exponential element-wise.

Parameters x (obs) – input observable

**Returns** the exponential of the input observable, element-wise.

## **Examples**

```
>>> expA = pyobs.exp(obsA)
```

## 3.2.3 pyobs.cosh

pyobs.cosh(x)

Return the Hyperbolic cosine element-wise.

 ${\bf Parameters} \ {\tt x} \ ({\tt obs}) - {\tt input} \ {\tt observable}$ 

Returns the hyperbolic cosine of the input observable, element-wise.

Return type obs

## **Examples**

```
>>> B = pyobs.cosh(obsA)
```

## 3.2.4 pyobs.sinh

pyobs.sinh(x)

Return the Hyperbolic sine element-wise.

 ${f Parameters} \ {\tt x} \ ({\tt obs}) - {\tt input} \ {\tt observable}$ 

Returns the hyperbolic sine of the input observable, element-wise.

Return type obs

## **Examples**

```
>>> B = pyobs.sinh(obsA)
```

## 3.2.5 pyobs.arccosh

```
pyobs.arccosh(x)
```

Return the inverse Hyperbolic cosine element-wise.

 ${\bf Parameters} \ {\tt x} \ ({\tt obs}) - {\tt input} \ {\tt observable}$ 

Returns the inverse hyperbolic cosine of the input observable, element-wise.

Return type obs

```
>>> B = pyobs.arccosh(obsA)
```

# 3.3 Special functions

besselk(v, x)

Modified Bessel function of the second kind of real order v, element-wise.

## 3.3.1 pyobs.besselk

pyobs.besselk(v, x)

Modified Bessel function of the second kind of real order v, element-wise.

#### **Parameters**

- v (float) order of the Bessel function
- $\bullet$  x (obs) real observable where to evaluate the Bessel function

**Returns** the modified bessel function computed for the input observable

## LINEAR ALGEBRA ROUTINES

```
pyobs.linalg.inv(x)
```

Compute the inverse of a square matrix

Parameters x (obs) - Matrix to be inverted

**Returns** (Multiplicative) inverse of x

Return type obs

#### **Examples**

```
>>> from pyobs.linalg import inv
>>> a = pyobs.obs()
>>> a.create('A',data,shape=(2,2))
>>> ainv = pyobs.inv(a)
```

#### **Notes**

If the number of dimensions is bigger than 2, x is treated as a stack of matrices residing in the last two indexes and broadcast accordingly.

## pyobs.linalg.eig(x)

Computes the eigenvalues and eigenvectors of a square matrix observable. The central values are computed using the *numpy.linalg.eig* routine.

Parameters x (obs) – a symmetric square matrix (observable) with dimensions NxN

**Returns** a vector observable with the eigenvalues and a matrix observable whose columns correspond to the eigenvectors

Return type list of obs

## Notes

The error on the eigenvectors is based on the assumption that the input matrix is symmetric. If this not respected, the returned eigenvectors will have under or over-estimated errors.

## **Examples**

```
>>> [w,v] = pyobs.linalg.eig(mat)
>>> for i in range(N):
>>>  # check eigenvalue equation
>>> print(mat @ v[:,i] - v[:,i] * w[i])
```

#### pyobs.linalg.eigLR(x)

Computes the eigenvalues and the left and right eigenvectors of a square matrix observable. The central values are computed using the numpy.linalg.eig routine.

**Parameters** x (obs) – a square matrix (observable) with dimensions NxN;

**Returns** a vector observable with the eigenvalues and two matrix observables whose columns correspond to the right and left eigenvectors respectively.

Return type list of obs

#### **Notes**

This input matrix is not expected to be symmetric. If it is the usage of *eig* is recommended for better performance.

#### **Examples**

```
>>> [1,v,w] = pyobs.linalg.eigLR(mat)
>>> for i in range(N):
>>>  # check eigenvalue equation
>>> print(mat @ v[:,i] - v[:,i] * 1[i])
>>> print(w[:,i] @ mat - w[:,i] * 1[i])
```

#### pyobs.linalg.matrix\_power(x, a)

Raises a square symmetric matrix to any non-integer power a.

### Parameters

- x (obs) a symmetric square matrix (observable) with dimensions NxN
- $\bullet$  a (float) the power

**Returns** the matrix raised to the power a

Return type obs

#### Notes

The calculation is based on the eigenvalue decomposition of the matrix.

```
>>> matsq = pyobs.linalg.matrix_power(mat, 2) # identical to mat @ mat
>>> matsqrt = pyobs.linalg.matrix_power(mat, -0.5)
>>> matsqrt @ mat @ matsqrt # return the identity
```

## THE MFIT CLASS

class pyobs.mfit(x, W, f, df, v='x')

Class to perform fits to multiple observables, via the minimization of the  $\chi^2$  function

$$\chi^2 = r^T W r$$
 ,  $r_i = y_i - \phi(\{p\}, \{x_i\})$ 

with  $\phi$  the model function,  $p_{\alpha}$  the model parameters  $(\alpha = 1, ..., N_{\alpha})$  and  $x_i^{\mu}$  the kinematic coordinates (i = 1, ..., N corresponds to the i-th kinematical point, and  $\mu$  labels the dimensions). The matrix W defines the metric of the  $\chi^2$  function.

The ideal choice for the matrix W is the inverse of the covariance matrix, which defines a correlated fit. Instead a more practical choice for cases where a reliable estimate of covariances is not available, is given by the inverse of the diagonal part of the covariance matrix, which instead defines an uncorrelated fit.

#### **Parameters**

- x (array) array with the values of the x-axis; 2-D arrays are accepted and the second dimension is interpreted with the index  $\mu$
- W(array) 2-D array with the matrix W; if a 1-D array is passed, the program inteprets it as the inverse of the diagonal entries of the covariance matrix
- f (function) callable function or lambda function defining  $\phi$ ; the program assumes  $x_i^{\mu}$  correspond to the first arguments
- df (function) callable function or lambda function returning an array that contains the gradient of  $\phi$ , namely  $\partial \phi(\{p\}, \{x_i\})/\partial p_{\alpha}$
- v(str, optional) a string with the list of variables used in f as the kinematic coordinates. Default value corresponds to x, which implies that f must be defined using x as first and unique kinematic variable.

## Notes

Once created the class must be called with at least one argument given by the observables corresponding to the data points  $y_i$ . See examples below.

```
>>> xax=[1,2,3,4]
>>> f=lambda x,p0,p1: p0 + p1*x
>>> df=lambda x,p0,p1: [1, x]
>>> [y,dy] = yobs1.error()
>>> W=1./dy**2
>>> fit1 = mfit(xax,W,f,df)
>>> pars = fit1(yobs1)
>>> print(pars)
0.925(35) 2.050(19)
```

eval(xax, pars)

Evaluates the function on a list of coordinates using the parameters obtained from a  $\chi^2$  minimization.

#### **Parameters**

- $xax(array, list\ of\ arrays)$  the coordinates  $x_i^{\mu}$  where the function must be evaluated. For combined fits, a list of arrays must be passed, one for each fit.
- pars (obs) the observable returned by calling this class

#### Returns

**observables corresponding to the functions evaluated** at the coordinates xax.

Return type obs, list of obs

## **Examples**

pars()

Prints the list of parameters

## 5.1 Additional functionalities

```
pyobs.symbolic.diff(f, x, dx)
```

Utility function to compute the gradient and hessian of a function using symbolic calculus

#### **Parameters**

- f (string) the reference function
- $\bullet$  x (string) variables that are not differentiated; different variables must be separated by a comma
- dx (string) variables that are differentiated; different variables must be separated by a comma

**Returns** (scalar) function f lambda : (vector) function of the gradient of f lambda : (matrix) function of the hessian of f

Return type lambda

#### **Notes**

The symbolic manipulation is based on the library sympy and the user must follow the sympy syntax when passing the argument f. The analytic form of the gradient and hessian can be printed by activating the 'diff' verbose flag.

```
>>> res = diff('a+b*x','x','a,b') # differentiate wrt to a and b
a + b*x
[1, x]
[[0, 0], [0, 0]]
>>> for i in range(3):
>>> print(res[i](0.4,1.,2.))
1.8
[1, 0.4]
[[0, 0], [0, 0]]
```

## ADDITIONAL UTILITY FUNCTIONS

pyobs.random.acrand(mu, sigma, tau, N)

Create synthetic autocorrelated Monte Carlo data

#### **Parameters**

- mu (int or float) the central value
- sigma (float) the square root of the variance of the observable in absence of autocorrelations
- tau (float) the integrated autocorrelation time
- N (int) the number of configurations/measurements

Returns the synthetic data

Return type list

**Note:** The expected error from a proper autocorrelation analysis is the product of sigma with the square root of tau divided by N.

## **Examples**

```
>>> data = pyobs.random.acrand(0.1234,0.0001,4.0,1000)
>>> obs = pyobs.obs(desc='test-acrand')
>>> obs.create('A',data)
>>> print(obs)
0.12341(26)
```

pyobs.random.acrandn(mu, cov, tau, N)

Create synthetic correlated Monte Carlo 1-D data

#### Parameters

- mu (*list of array*) the central values of corresponding observable; a 1-D array is expected
- cov (array) the covariance matrix of the observable (in absence of autocorrelations); if a 1-D array is passed, a diagonal covariance matrix is assumed
- tau (float) the integrated autocorrelation time
- $\mathbb{N}(int)$  the number of configurations/measurements

## Returns

**2-D** array with the synthetic data, such that each row corresponds to a configuration

Return type numpy.ndarray

## **CHAPTER**

# **SEVEN**

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# **PYTHON MODULE INDEX**

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