- >>> import numpy as np
- >>> import scipy as sp

% ipython notebook --pylab inline

### Scientific Programming I



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See also: Nat Butler's lecture notes from last year's bootcamp

## SciPy packages

This lecture explores:

linalg, fftpack, optimize, integrate and interpolate

#### SciPy: numerical algorithms galore

- linalg: Linear algebra routines (including BLAS/LAPACK)
- sparse : Sparse Matrices (including UMFPACK, ARPACK,...)
- fftpack : Discrete Fourier Transform algorithms
- cluster: Vector Quantization / Kmeans
- odr : Orthogonal Distance Regression
- special: Special Functions (Airy, Bessel, etc).
- stats: Statistical Functions
- optimize : Optimization Tools
- maxentropy: Routines for fitting maximum entropy models
- integrate : Numerical Integration routines
- ndimage : n-dimensional image package
- interpolate : Interpolation Tools
- signal : Signal Processing Tools
- io : Data input and output



### Overview of SciPy challenges in this lecture

 Generating random variables with the same distribution as an input data set Using: sp.integrate, sp.interpolate, np.random
 Presupposes: Some statistics background, but I will provide a primer

 Calculating a Fourier transform of 1D data, with error bars Using: sp.fftpack, np.random
 Presupposes: Knowledge of Fourier transformation

Fitting a model to (perhaps covariant) data
 Using: sp.optimize, sp.linalg
 Presupposes: A bit more statistics background, which, again, I will describe

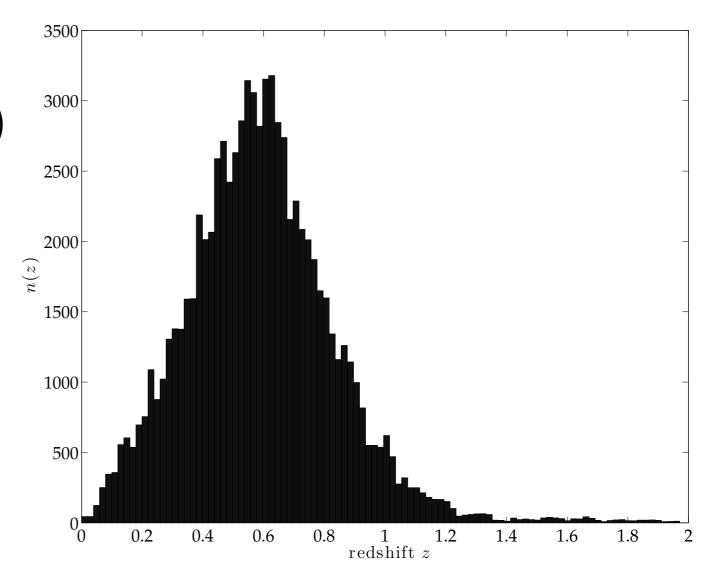
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scipy.package_name.[tab]
scipy.package_name.function_name?
http://docs.scipy.org/doc/scipy/reference/ SciPy reference guide, tutorial
```

- I. Integration and interpolation
- >>> from scipy.integrate import \*
- >>> from scipy.interpolate import \*

### Problem: generating random variables from data

Data: [0.674, 1.053, 0.453...]

PDF:  $\hat{p}(z) \propto \text{hist}(\text{Data})$ 

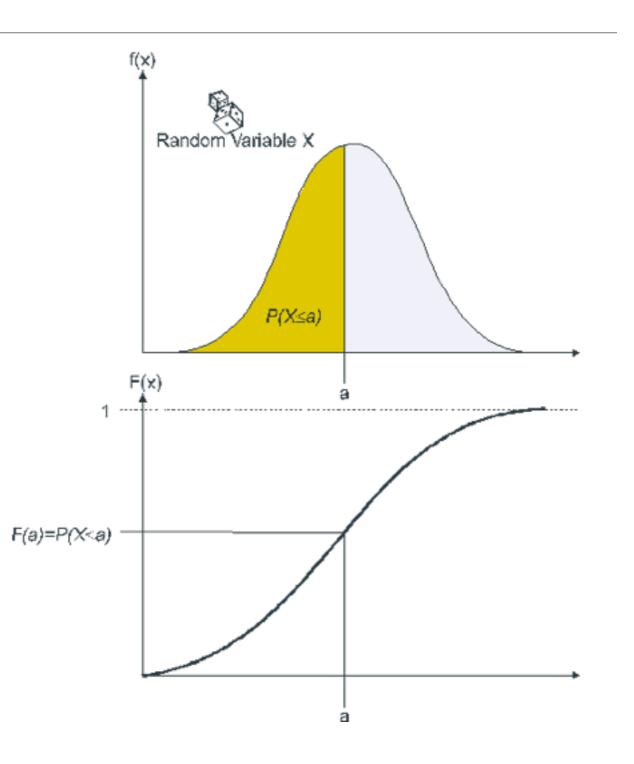


### Problem: generating random variables from data

Data: [0.674, 1.053, 0.453...]

PDF:  $\hat{p}(z) \propto \text{hist}(\text{Data})$ 

 $CDF : \hat{P}(z) = \int_0^z p(z')dz'$ 



### Problem: generating random variables from data

Data:  $[0.674, 1.053, 0.453...]^{1.8}$ 1.2 PDF:  $\hat{p}(z) \propto \text{hist}(\text{Data})$ 0.8  $CDF : \hat{P}(z) = \int_{0}^{z} p(z')dz'$ 0.5 1.5

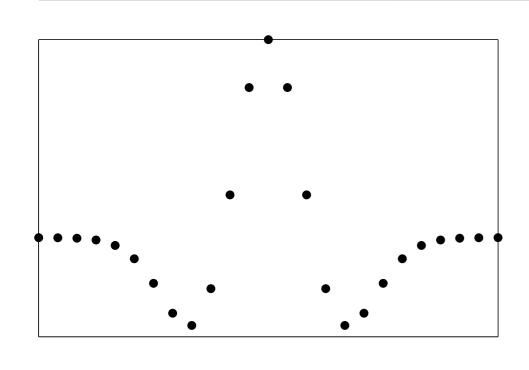
If  $U \sim \text{Uniform}[0, 1)$ , then  $\hat{P}^{-1}(U) \sim \text{Data}$ 



#### II. An uncertain Fourier transform

- >>> from numpy.random import \*
- >>> from scipy.fftpack import \*

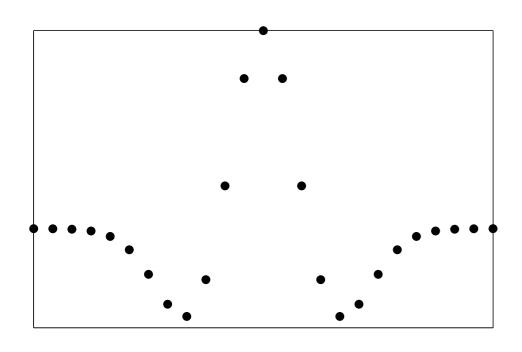
### Problem: FFT with error bars

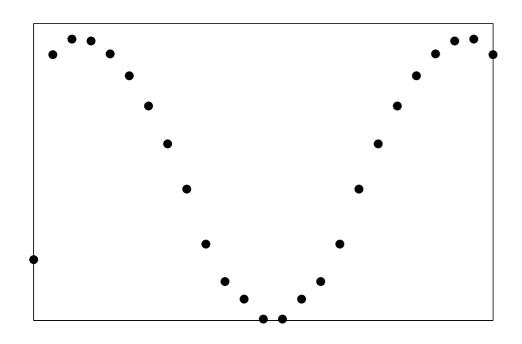


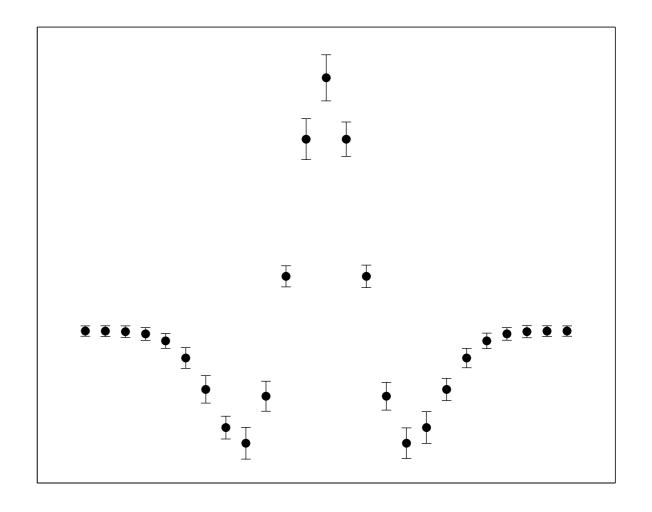
$$F(\tau) = \mathcal{F}(f(t))$$

$$P(\tau) = F(\tau) \times F^*(\tau)$$

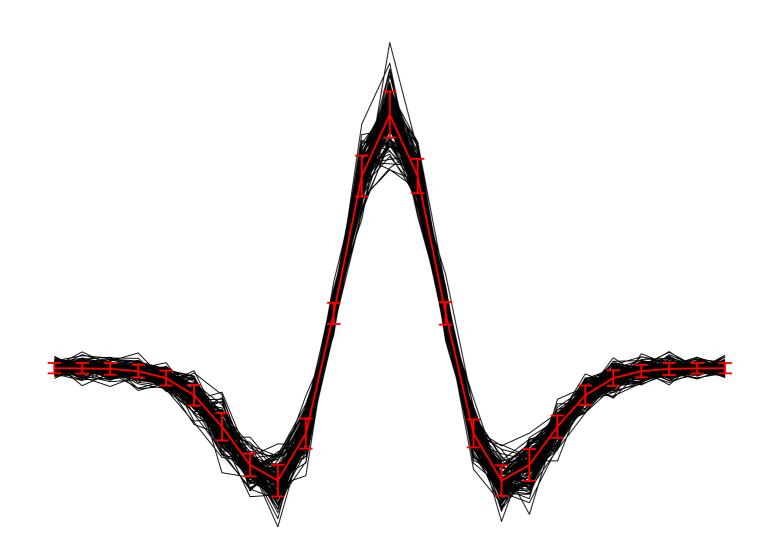
## Problem: FFT with error bars







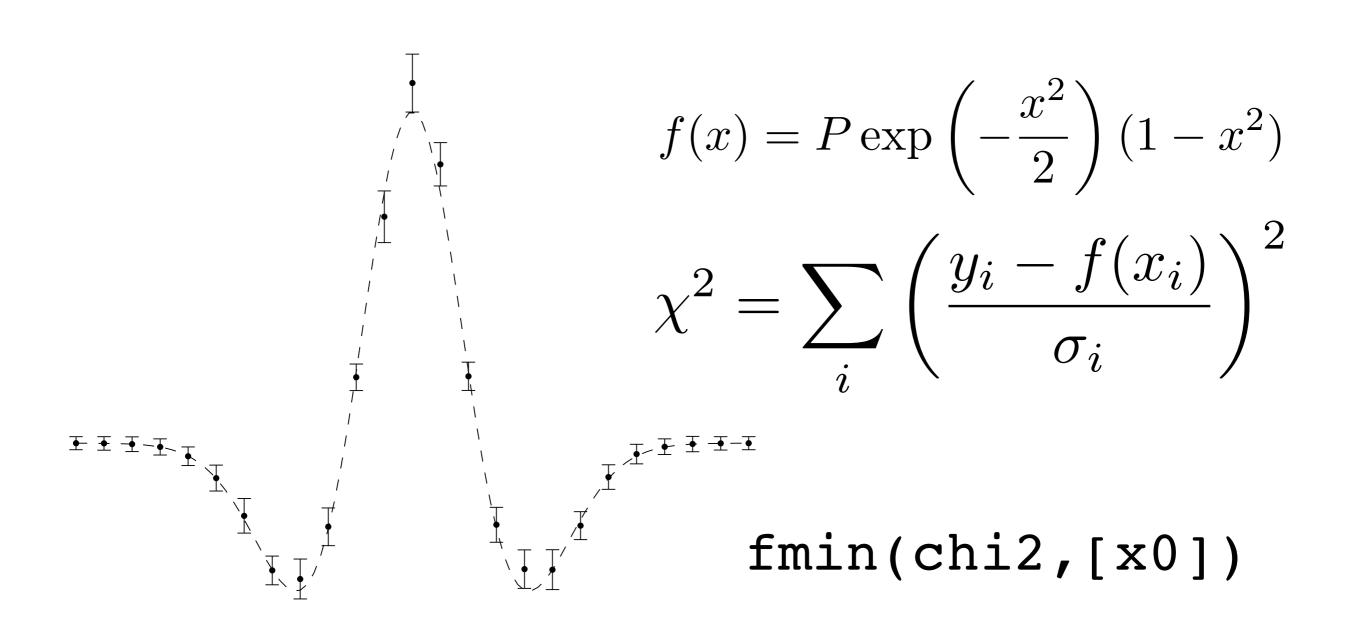
# (Unsatisfactory) solution: brute sampling



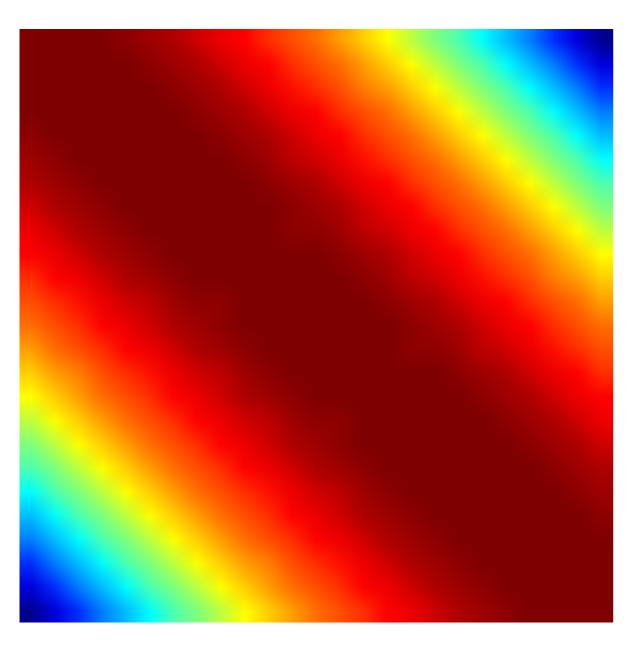


- III. Fitting (likelihood estimation) with covariant data
- >>> import scipy.linalg as la
- >>> import scipy.optimize as opt

### Problem: non-linear model fitting



### Problem: non-linear model fitting with covariance



$$f(x) = P \exp\left(-\frac{x^2}{2}\right) (1 - x^2)$$
$$\Delta = y_i - f(x_i)$$
$$\chi^2 = \Delta \mathbf{C}^{-1} \Delta'$$

fmin(chi2,[x0])

### Eigendecomposition (for covariant data points)

- For a symmetric matrix, eigendecomposition can be used to deal with tricky inversions
- One method is to locate very small eigenvalues, and set their inverse to zero.

$$C = VEV' \Rightarrow C^{-1} = VE^{-1}V'$$