

# Data analysis exercises

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please copy the directory

[/afs/ifh.de/group/ape/lap/12/analysis/octave\\_routines](http://afs/ifh.de/group/ape/lap/12/analysis/octave_routines)

please start octave either locally or on pax8x

[/afs/ifh.de/group/ape/lap/12/octave](http://afs/ifh.de/group/ape/lap/12/octave)

# Octave basics

Octave is an interpreter:

```
octave:1> a=1  
a = 1
```

Don't print results:

```
octave:1> a=1;
```

Print an expression:

```
octave:1> a  
a = 1
```

# Octave arrays

Octave can handle arrays well:

```
octave:1> a=[1,2,3,4,5]
a =
    1    2    3    4    5
```

Simpler:

```
octave:1> a=1:5
a =
    1    2    3    4    5
```

Also with step sizes:

```
octave:1> a=5:-0.5:3
a =
    5.0000    4.5000    4.0000    3.5000    3.0000
```

Octave treats objects as matrices:

```
octave:1> a=[1,1;0,2]
```

```
a =
```

1	1
0	2

Arithmetic operations are natively matrix:

```
octave:1> a^2
```

```
ans =
```

1	3
0	4

If you want it element-wise:

```
octave:1> a.^2
```

```
ans =
```

1	1
0	4

# Octave vectors

Octave octave distinguished between row vectors

```
octave:1> a=[1,2]
a =
    1    2
```

and column vectors

```
octave:1> b=[1;2]
b =
    1
    2
```

Transposition is simple:

```
octave:1> b'
ans =
    1    2
```

# Octave matrix multiply

If we define a row vector

```
octave:1> a=[1, 2]
a =
 1   2
```

there is of course a difference between

```
octave:1> a*a'
ans = 5
```

and

```
octave:1> a'*a
ans =
 1   2
 2   4
```

# Octave eigenvalues

Let us define a symmetric matrix:

```
octave:1> m=[2,1;1,4]
m =
 2   1
 1   4
```

We can find the eigenvalues:

```
octave:1> eig(m)
ans =
 1.5858
 4.4142
```

# Octave example

To find the eigenvectors as well, use:

```
octave:1> [vec, val]=eig(m)
vec =
-0.92388    0.38268
 0.38268    0.92388
val =
Diagonal Matrix
 1.5858        0
      0    4.4142
```

So we can reconstruct the original matrix:

```
octave:1> vec*val*vec'
ans =
 2.00000    1.00000
 1.00000    4.00000
```

# Octave matrix functions

Initializing matrix (zeros, rand, diag) example:

```
octave:1> m=rand(2)  
m =  
0.089507    0.047387  
0.718905    0.878561
```

Sum (and prod) reduces one dimension:

```
octave:1> sum(m)  
ans =  
0.80841    0.92595  
octave:2> sum(m, 2)  
ans =  
0.13689  
1.59747
```

# Octave cell arrays

One last detail: cell arrays

```
octave:1> a={1, "string"}  
a =  
{  
    [1,1] = 1  
    [1,2] = string  
}
```

They may contain different objects and are accessed as:

```
octave:1> a{1}  
ans = 1  
octave:2> a{2}  
ans = string
```

# Octave cell arrays

Cell arrays can easily be enlarged:

```
octave:1> a=[a [1,2]]
```

```
a =
```

```
{
```

```
[1,1] = 1
```

```
[1,2] = string
```

```
[1,3] =
```

```
1 2
```

```
}
```

```
octave:2>a{3}
```

```
ans =
```

```
1 2
```

## Introduction

# Let's start!

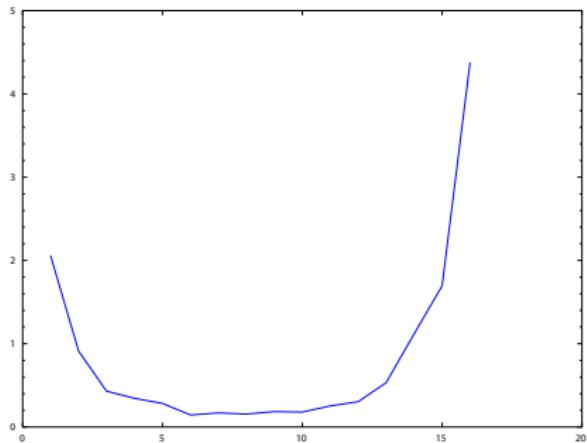
Please open the file “readme.m”

Produce one fake “pion” propagator:

```
octave:> myprop1=piprop(16)
```

and plot it:

```
octave:1> plot(myprop1{1})
```



directory:

/afs/ifh.de/group/ape  
/lap/12/analysis  
/octave\_routines

## Displaying data

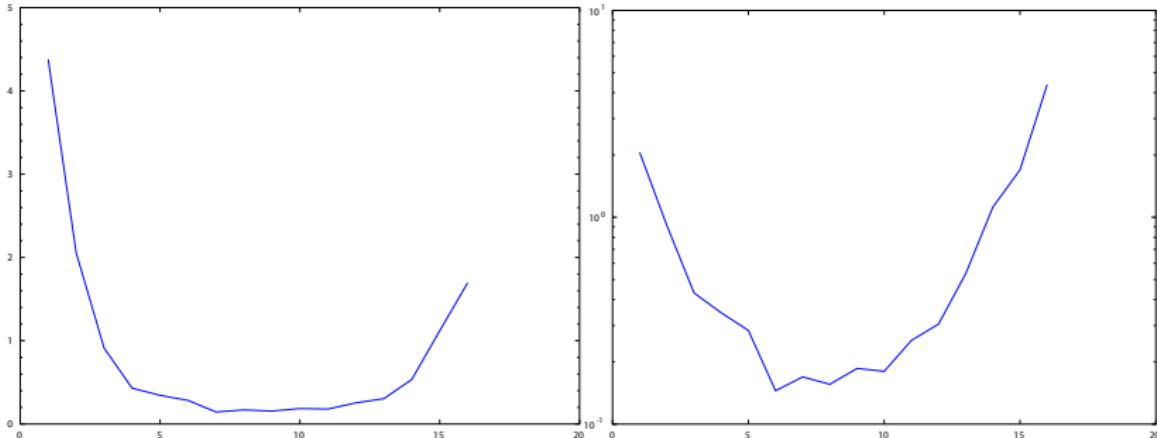
Please don't look at "piprop.m" or any "[..]prop.m" file yet!

If you like the origin at 0 instead of 16:

```
octave:1> plot(myprop1{1}([end 1:end-1]))
```

For a logarithmic y-scale plot:

```
octave:1> semilogy(myprop1{1})
```



## Introduction

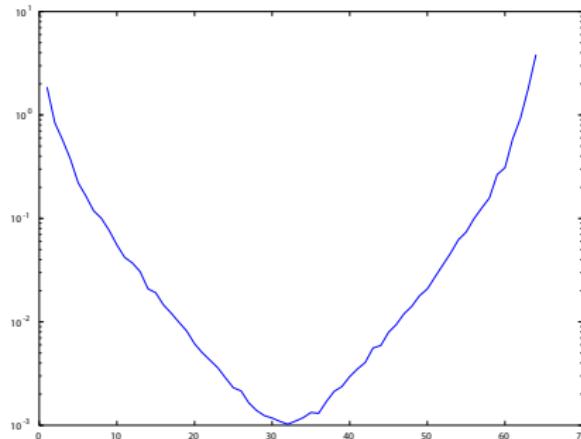
# The first ensemble

Now 100 configs with  $N_T = 64$ :

```
octave:1> prop=piprop(64,100);
```

For a logarithmic y-scale plot:

```
octave:1> semilogy(prop{1})
```



## Introduction

# Bootstrapping

Produce 200 bootstrap samples:

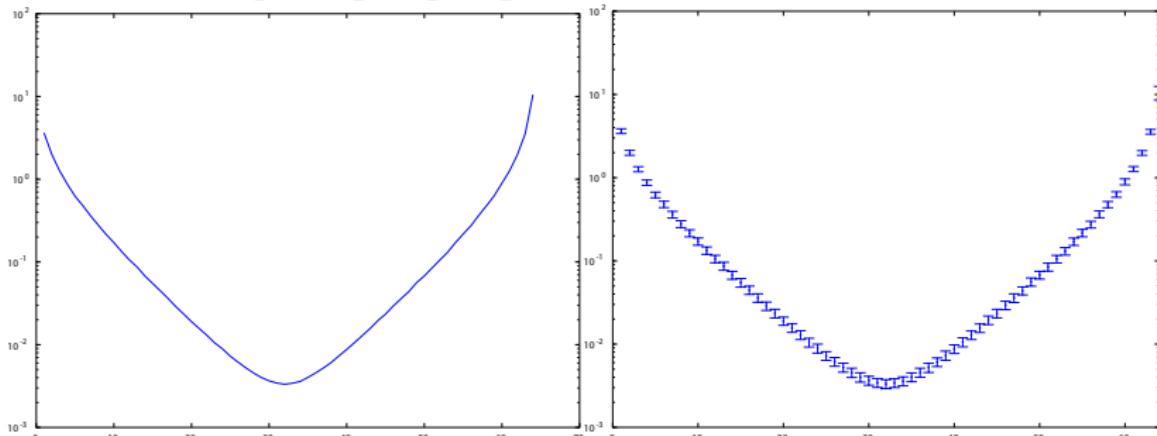
```
octave:1> bp=boot (prop, 200);
```

Plot the average (column 201):

```
octave:1> semilogy (bp (:, end) )
```

Routine for plotting with errors:

```
octave:1> plotprop (bp)
```



## Introduction

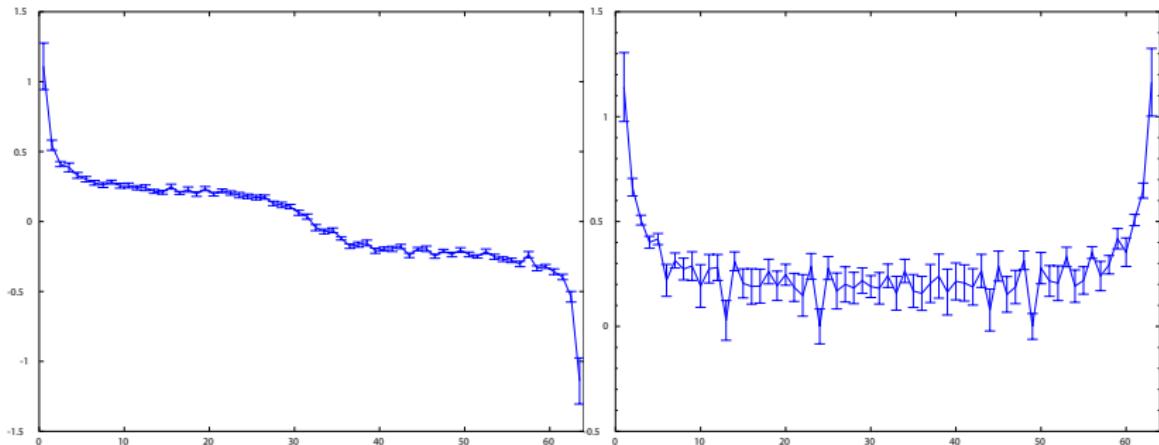
## Effective mass

Effective mass with simple exp:

```
octave:1> plotplat(bp)
```

Effective mass with cosh:

```
octave:1> plotsympat(bp)
```



## Introduction

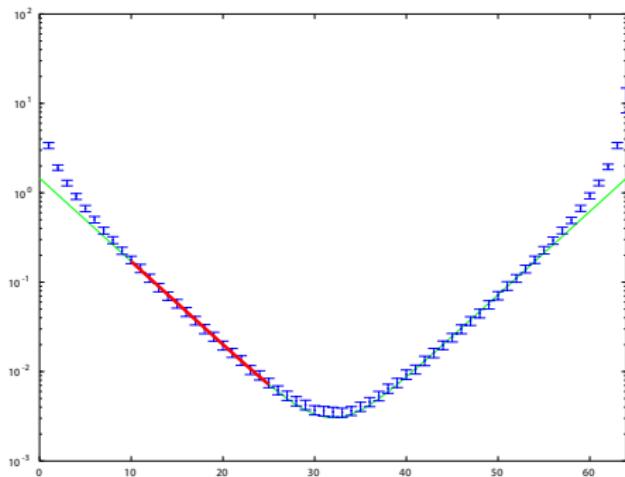
## Mass fit

```
uncorrelated_fit=0  
fully_correlated_fit=-1
```

```
central_value_only=1  
full_bootstrap_fit=0
```

```
do_plot=1  
dont_plot=0
```

```
massfit(bp, uncorrelated_fit, 10, 25,  
        central_value_only, do_plot)
```



## Introduction

# Multiple output variables

```
octave:1> [m,em,c,ec,fitq]=massfit(bp,0,10,25)
m = 0.21448
em = 0.0024118
c = 1.4575
ec = 0.13362
fitq = 0.99984
```

## Introduction

## Excited state fit

You might want to try a fit with 2 coshs:

```
octave:1> [m,em,c,ec,fitq,m2,em2,c2,ec2]=  
exsfhit(bp,0,5,25,0,1)
```

m = 0.20403

em = 0.0028877

c = 1.1466

ec = 0.15216

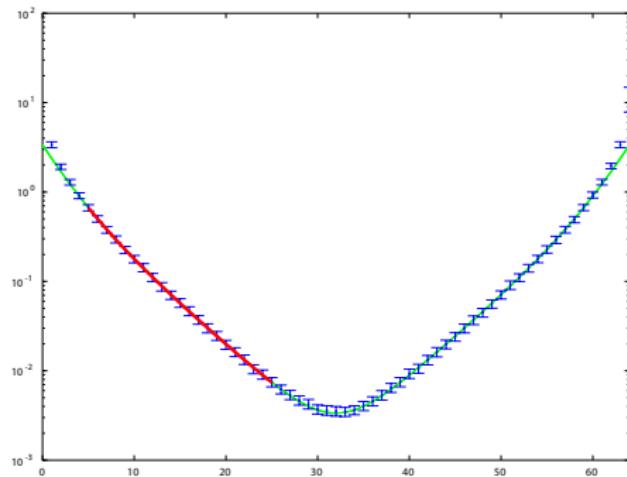
fitq = 1.00000

m2 = 0.43349

em2 = 0.026877

c2 = 2.2185

ec2 = 0.28166



## Correlation matrix

## Correlation matrix eigenvalues

## Eigenvalues of correlation matrix:

```
octave:1> correlation_matrix_solution(bp(10:25, :))  
ans =  
1.1263e-03  
1.5686e-03  
1.8096e-03  
2.8204e-03  
3.8986e-03  
7.4055e-03  
8.9888e-03  
1.2718e-02  
1.6033e-02  
2.2189e-02  
2.8986e-02  
3.4987e-02
```

## Correlation matrix

## Your routine

Eigenvalues of correlation matrix:

```
function ev=correlation_matrix_template(bp)
# this is a template for a routine that should
# compute the normalized correation matrix and
# its eigenvalues from a bootstrap array
N=size(bp)(1);      # number of points
NB=size(bp)(2)-1;   # number of bootstrap samples

# insert code here

    ev=eig(cor);
endfunction
```

## Correlation matrix

## Correlation matrix solution

```

# take out normalization
bav=sum(bp(:,1:N),2)/NB;
ac=sqrt((sum(bp(:,1:N).^2,2)/NB-bav.^2) *
(1+(1/(NB-1))));

# form the covariance matrix
cor=zeros(N); # initialize cor as N*N matrix
for x=1:N
    for y=1:N
        cor(y,x)=(bp(y,1:N)-bav(y))* 
            (bp(x,1:N)-bav(x))' / 
            (NB-1)/(ac(x)*ac(y));
    end
end

```

# Find the mass!

- Explore the fitting routine
- Try varying nfonf, nboot, NT, fit range, correlation
- Find the “pion” mass and its error
- Upload your results to

/afs/ifh.de/group/ape/lap/12/analysis/mpi\_results/xxxx\_mpi

/afs/ifh.de/group/ape/lap/12/analysis/mpi\_results/xxxx\_err

- If you are done, you can try “rhoprop” instead of “piprop”

have fun!

## Finding

## Going to the physical point

Please open the file “day2.m”  
generate one well behaved ensemble at “bare quark mass” mq:

NT=64  
NBOOT=100  
NCONF=100  
mq=-0.01

octave:1> prp=easy\_piprop(NT,mq,NCONF);

Bootstrap it and look at the pion mass:

octave:1> bpr=boot(prp,NBOOT);

octave:2> m=massfit(bpr,0,15,30,1,1)

Only central value computed!

m = 0.57282

## Finding

# The physical point

We are at  $1/a = 1.5 \text{ GeV}$ , so the target “physical”  $m_\pi a = 0.09$

Produce ensembles that allow you to go to the physical point

- You can either interpolate or extrapolate
- Neat the “physical point” propagators are more noisy

## Finding

## Generating ensembles

Template ensemble generation:

```
# a random table of quark masses
masstab=[0,-0.04,-0.06,-0.08]

for i=1:length(masstab)
    prp=easy_piprop(NT,masstab(i),NCONF);
    bp{i}=boot(prp,NBOOT);
# mb and cb are bootstrap results for m and c
    [m,em,c,ec,q,mb{i},cb{i}]=massfit(bp{i},0,15,30);
# extract f from c=f^2*m^2
    fb{i}=sqrt(abs(cb{i}))/mb{i};
end
```

## Extra/Interpolation

## Going to the physical point

Please open the file “chiralfit\_template.m”

This is a long fit routine with a very simple core:

```
stp=[d0,k0];  
# do the constrained fit  
[par,ch2c,info]=sqp(stp',@constrainedfit,  
[],[],[],[],1000);
```

A  $\chi^2$  function is minimized, x-errors are yet ignored

## Extra/Interpolation

## Function to minimize

```
function chisq=constrainedfit(par)
# chi^2 function for constrained fpi-mpi^2 fit
global _f _m _er _n

chisq=0;
for i=1:_n
    mm=_m(i);
    ff=fps(mm,par);
    ex=[_m(i)-mm,_f(i)-ff];
    chisq+=ex*_er{i}*ex';
end
chisq;
endfunction
```

## Extra/Interpolation

## Fit function

x-values (\_m), y-values (\_f) and the inverse covariance matrix (\_er) are global:

```
global _f _m _er
```

```
octave:1> global _f
octave:2> _f
_f =
 0.068717    0.072484    0.075524    0.073040
```

The actual linear fit function:

```
function ff=fps(m,c)
# the actual fit function
ff=c(1)+c(2)*m;
endfunction
```

## Extra/Interpolation

## Full fit exercise

Then all masses are added as parameters:

```
stp=par';  
for i=1:N  
    stp=[stp _m(i)];  
end
```

And the full fit is performed:

```
[par, ch2, info]=sqp(stp', @fullfit, [],[],[],[],1000);
```

Complete the routine fullit!

The solution is in “chiralfit\_solution.m”

## Extra/Interpolation

## Completing full fit

```
function chisq=fullfit(par)
# chi^2 function for full fpi-mpi^2 fit
global _f _m _er _n

#####
# YOUR CODE GOES HERE
# remember: par(3:) are
# the fitted masses
# (x-axis)
#####
endfunction
```

The solution is in “chiralfit\_solution.m”

## Extra/Interpolation

```
function chisq=fullfit(par)
# chi^2 function for full fpi-mpi^2 fit
global _f _m _er _n

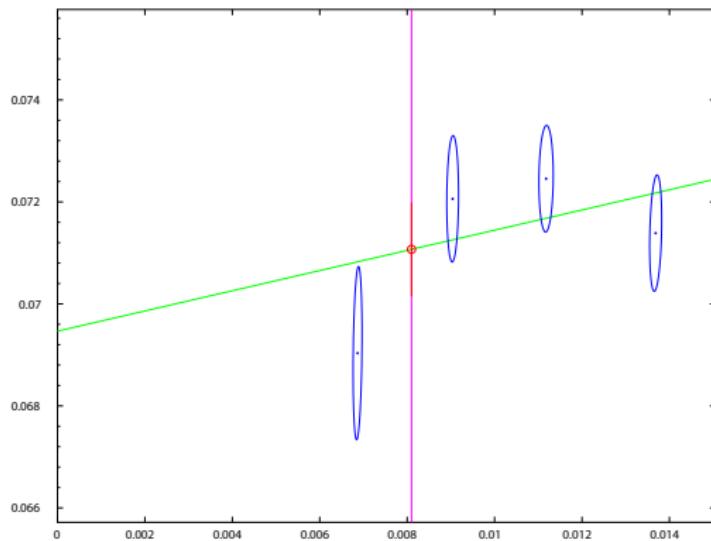
npar=2;
chisq=0;
for i=1:_n
    mm=par(npar+i);
    ff=fps(mm,par(1:npar));
    ex=[_m(i)-mm,_f(i)-ff];
    chisq+=ex*_er{i}*ex';
end
chisq;
endfunction
```

Extra/Interpolation

# Doing the fit

Using the chiral fit:

```
> [fpi,efpi,q,qc]=chiralfit_solution(fb,mb,-1,0,1)
fpi = 0.071069
efpi = 9.1972e-04
q = 0.51122
qc = 0.49685
```



## Extra/Interpolation

## Different fit forms

Setting the x-axis as  $m^2$  (line 33):

```
# use mpi^2 as x-axis
mb{ i }=mb{ i }.^2;
```

Vary the 2-parameter fit form (end of file):

```
function ff=fps(m,c)
# the actual fit function
ff=c(1)+c(2)*m;
endfunction
```

Do a 3-parameter fit:

```
[fpi,epi,q,qc]=chiralfit_quadratic(fb,mb,-1,1,1)
```

Extra/Interpolation

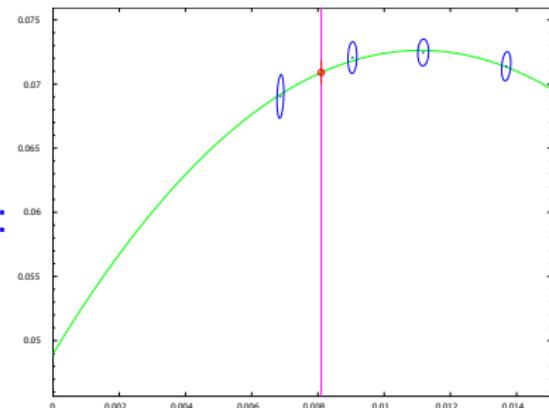
# Quadratic fit

Using the chiral fit:

```
> [fpi,efpi,q,qc]=chiralfit_quadratic(fb,mb,-1,0,1)
fpi = 0.070900
efpi = 9.0830e-04
q = 0.97827
qc = 0.97816
```

Vary the 3-parameter fit form (end of file):

```
function ff=fps(m,c)
# the actual fit function
    ff=c(1)+c(2)*m+c(3)*m.^2;
endfunction
```



## Extra/Interpolation

## Find f at the physical point!

- Explore the fitting routines - change fit forms
- Try varying ensembles and fit parameters
- Find the “decay constant”  $f$  and its error
- Upload your results to

/afs/ifh.de/group/ape/lap/12/analysis/fpi\_results/xxxx\_fpi

/afs/ifh.de/group/ape/lap/12/analysis/fpi\_results/xxxx\_err

- If you are done, try estimating the systematics
- If you want the solution, look at “cheat.m” and “physpoint.m”

have fun!