

Matlab tutorial, 02.05.06 & 03.05.06

Vectors and matrices

```
Y=[1 2 3]
Y = [1 2 3];
Z=Y'                                % transpose of matrix Y
Z*Y
Y*Z
Y.*Y
Y.^2
Y^2
W=Z*Y
W^2
W.^2
W(1,1)=2; W(2,2)=4; W(3,3)=8; % change elements of matrix W manually
W
det(W)
inv(W)
B=2*ones(3,1)                      % ones(n,m) creates an (n x m) - identity matrix
X=W\B                               % solves the linear equation WX=B
inv(W)*B
W=W(1:2,:)                          % choose the first and second row of matrix W
B=B(1:2)
Y=W\B                               % if W is not a square matrix, the command solves a least-squares
                                   % problem, i.e. it minimizes  $\|WY - B\|^2 = \min$  - compare MATLAB help
                                   % for details.
```

Loading and saving data

```
load BASFR
save TEMP BASFR
```

Here, the data BASFR are saved in a file TEMP.MAT; you can save more than one variable in the same file by using the command save TEMP BASFR BMW DAXR ...

The command save TEMP saves all variables in the workspace in the file TEMP.MAT.

Plotting data

```
plot(BASFR)
plot(BASFR,'*r')
```

You can manipulate a figure by using the Edit commands of the window in which the figure is displayed. Just play around with the possibilities. Using the File commands, you can save the figure - e.g. as file BASFR.FIG - in MATLAB's own graphic format. You can open and manipulate it later. You also can export a figure into some other graphics format for insertion into a paper.

```
t=1:1:746;
t=t';
plot(t,BASFR)
t=t.^2;
plot(t,BASFR)
load DAXR;
plot(t,BASFR,t,DAXR)
plot(t,[BASFR DAXR]) % the column vectors BASFR and DAXR are merged to a matrix with 2
                    % columns.
subplot(2,1,1)      % breaks the Figure window into a 2-by-1 matrix of small axes, refer to first
                    % axis.
plot(BASFR)
subplot(2,1,2)      % refer to second axis
plot(DAXR)
```

There are also SURF and MESH commands for the 3D-plot. Check it in the MATLAB help.

Plotting functions

```
x=-3.5:0.05:3.5;
p=exp(-x.^2/2)/sqrt(2*pi); % Is it familiar to you?
plot(p)
plot(x,p)
f=normpdf(x,0,1); % Returns the normal probability density function with mean 0, and
                  % standard deviation 1, at the values in x
hold on % Keep the previous plot .
plot(x,f)
f2=normpdf(x,1,0.5);
plot(x,[f;f2])
```

Generating random variables

```
z=normrnd(0,1,200,1); % Generate N(0,1)-distributed random variables
m=mean(z)
s=std(z)
normplot(z) % It is linear if z is normally distributed
help normplot
hist(z,15)
hist(z,21)
histfit(z,21) % Fit a normal density function to the histogram
```

Kernel estimate of probability density

The following commands calculate an estimate \hat{p} of the probability density p of a sample of iid random variables (in vector z - in our case standard normal). The theory is developed in the lecture on nonparametric statistics. At the moment, you have to believe that it is consistent if the sample size n goes to infinity and the smoothing parameter h goes to 0 such that nh goes to infinity.

```
n=length(z)
nx=length(x);
h=0.2;
pe=p;
for i=1:nx
u=normpdf((x(i)-z)/h); pe(i)=sum(u)/(n*h); end
plot(x,[p;pe]) % plot of the true density and the estimate - repeat the calculation with
               % different values of h and look what happens.
pe2=kernpde(z,x,0.1);
plot(x,[p;pe;pe2])
```

As a second example, consider a mixture of two normal distributions with true density p

```
p=0.8*normpdf(x,0,0.5)+0.2*normpdf(x,2,0.2); % the corresponding random variables can be written
                                              % as  $B*Z_1+(1-B)*Z_2$  where  $Z_1, Z_2$  are variables from the
                                              % two normal laws and  $B$  is a Bernoulli, i.e. 0-1-variable.

plot(x,p)
B=ceil(unifrnd(0,1,200,1)-0.2); % generate B in the pedestrian way using the basic uniform
                               % random number generator
z=B.*normrnd(0,0.5,200,1)+(1-B).*normrnd(2,0.2,200,1);
pe=kernpde(z,x,0.1);
plot(x,[p;pe])
B=binornd(1,0.8,200,1); % alternatively generate B using the special random generators of the
                       % statistics toolbox
z=B.*normrnd(0,0.5,200,1)+(1-B).*normrnd(2,0.2,200,1);
pe=kernpde(z,x,0.1);
plot(x,[p;pe])
normplot(z)
```

Your task. write a function file **pdkernest.m** which compute the probability density estimator, which takes z , x , and h as input.

Distribution of binomial parameter estimate

We want to see how the generated binomial random variables distributed is. We also compare it with the normal distribution with the parameter p , s .

```
x=binornd(50,0.1,5000,1);
x=x/50;
mean(x)
std(x)
sqrt(0.1*0.9/50)           % compute the variance manually
median(x)
iqr(x)                       % compute the interquartile range
[min(x) max(x)]
hist(x,51)
hist(x,31)
histfit(x,31)
help boxplot
boxplot(x)
s=sqrt(0.9*0.1/50);
z=normrnd(0.1,s,5000,1);
boxplot(z)
boxplot([x z])
skewness(x)
skewness([x z])
```

What happens if you change 50 to 500? To 1000? Compare with the normal distribution!

Chi-square goodness-of-fit test

H_0 : x is normally distributed vs. H_1 : x has another distribution

Test statistics: $\chi^2 = \sum (z_j - e_j)^2 / e_j$

z_j = number of observations lie in interval j

e_j = expected number of observations lie in interval j

Reject H_0 if $\chi^2 >$ quantile of chi square distribution

```
intup=0.021:0.02:0.201;      % define the intervals
intup=[intup 0.3];
m=mean(x); s=std(x);
max(x)
s=zeros(11,1);
for i=1:11
s(i)=sum(ceil(x-intup(i))); end
s
z=zeros(11,1);
for i=2:11
z(i)=s(i-1)-s(i); end
z(1)=5000-s(1);
z
m=mean(x);
s=std(x);
e=normcdf(intup,m,s);        % computes the normal cdf with mean m and standard deviation s at
                              % the values in intup.
e(2:11)=e(2:11)-e(1:10);     % find the probability for an observation to be in interval j
e(1)=e(1)-normcdf(0,m,s);
e=5000*e';
[z e]
plot([z e], '*')
chi2=sum((z-e).^2./e);
chi2
chi2inv(0.95,8)
```

Is x normally distributed?