Tutorial 3⁸:

Statistics with MATLAB

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The purpose of this tutorial \dot{s} to present several statistics techniques using Matlab Statistics toolbox. For this tutorial, we assume that you know the basics of Matlab (covered in Tutorial 1) and the basics of statistics. The tutorial purpose is to teach you how to use the Matlab built-in functions to calculate the statistics for different data sets in different applications; the tutorial is intended for users running a professional version of MATLAB 6.5, Release 13.

Topics discussed in this tutorial include:

- 1. Descriptive statistics
- 2. Linear Models
- 3. Cluster analysis
- 4. Principal component analysis

1. Descriptive statistics

A. Correlation coefficient for two variables: "corrcoef.m"

>> a =

1 1 3 4

» b= [1 2 2 3]

b =

1 2 2 3

» corrcoef(a,b)

ans =

 $\begin{array}{rrr} 1.0000 & 0.8165 \\ 0.8165 & 1.0000 \end{array}$

Question: What large the value of 'ans' should be in order to obtain very strongly correlated variables?

B. *Standard deviation* of a variable: "std.m" >> » a=[3 5 7 8 9 11]

a =

3 5 7 8 9 11

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```
» std(a)
```

```
ans =
```

2.8577

C. Variance of a variable: "var.m"

>> » var(a)

ans =

8.1667

D. *Percentiles* for a data set: "prctile.m"

```
>> » prctile(a,50)
```

ans =

7.5000

```
» prctile(a,25)
```

ans =

```
5
```

```
» prctile(a,75)
```

ans =

```
9
```

```
» prctile(a,1)
```

ans =

3

```
» prctile(a,100)
```

```
ans =
```

11

E. *Range* for a data set: "range.m" >> range(a)

ans = 8

2. Linear models

Regression analysis: "regress.m"

>> b=[13 15 15 17 18 20]'

b = » a=[3 5 7 8 9 11]' a = » coef=regress(a,b) coef =

0.4516

3. Cluster analysis

Hierarchical clustering on a set of data: "clusterdata.m"

» X=[1 1 2 3;7 8 9 7;1 3 2 1;10 9 11 9;2 2 1 1;3 4 1 2]

 $\mathbf{X} =$

» T = clusterdata(X,0.8)

T =

T = CLUSTERDATA(X,CUTOFF) is the same as

Y = pdist(X,'euclid'); Z = linkage(Y, 'single'); T = cluster(Z, CUTOFF);

4. Principal Component Analysis (for dimensionality reduction)

Principal components analysis from raw data: "princomp.m"
» [pc, score, latent, tsquare]=princomp(X)

pc =

0.4961	-0.3697	0.6397	-0.4561
0.4316	-0.6706	-0.4003	0.4515
0.6032	0.4179	-0.5296	-0.4254
0.4514	0.4889	0.3874	0.6381

score =

-4.7824	2.0736	0.3947	0.2489
7.2433	0.0423	-0.7265	0.2473
-4.8221	-0.2454	-1.1806	-0.1243
11.2723	0.0763	0.5080	-0.2441
-5.3608	-0.3624	0.3889	-0.6064
-3.5502	-1.5843	0.6155	0.4786

latent =

53.3994 1.4018 0.5731 0.1590

tsquare =

4.1571 2.2893 3.0077 3.2088 3.2088 4.1284

5. Application

Apply correlation analysis, dimensionality reduction, and clustering techniques for the data containing the texture characterization for 193 CT images. The features/attributes consist of 11 texture descriptors calculated from the raw data and used to classify the tissues' textures of five human body organs: heart, backbone, liver, spleen, and kidneys. The data is contained in the file run_length_des.m".