

## Engineering Maths EMAT 30007, Parameter Estimation(1)

### 1 Estimation of population mean - confidence intervals

*The confidence intervals for the parameter  $\mu$  that we described in this section are all based on point estimates that could be assumed to be approximately normally distributed.*

- (a) To work out the coefficient in front of the standard error in the margin of error ( $\mu$  estimate  $\pm$  margin of error) you can use the commands `norminv` and `tinv` in the following code lines:

*In case of normal distribution:*

```
alpha=0.05; % for 95% confidence level
zv=norminv(1-alpha/2) % z-value for two-sided CI
zv=norminv(1-alpha) % z-value for one-sided CI
```

*In case of t-distribution:*

```
n=25; % sample size
tv=tinv(1-alpha/2, n-1) % t-value for two-sided CI
tv=tinv(1-alpha, n-1) % t-value for one-sided CI
```

- (b) The following measurement were recorded for the drying time (in hours) of a certain paint: 3.4, 2.5, 4.8, 2.9, 3.6, 2.8, 3.3, 5.6, 3.7, 2.8, 4.4, 4.0, 5.2, 3.0, 4.8. Assume that measurements represent a random sample from a normal distribution. Construct a two-sided 95% confidence interval for working out the mean drying time.

```
>> clear all
>> x=[3.4, 2.5, 4.8, 2.9, 3.6, 2.8, 3.3, 5.6, 3.7, 2.8, 4.4, 4.0, 5.2, 3.0, 4.8] % random samp
>> n=length(x);
>> mx=mean(x);
>> s=std(x);
>> alpha=0.05 ; % for 95% confidence
>> tcr=tinv(1-alpha/2, n-1); % critical t value
>> me=tcr*s/sqrt(n); % margin of error
>> CI1= mx-me; % lower CI bound
>> CI2= mx+me; % upper CI bound
```

- (c) Why was t-distribution used?
- (d) Explain what the phrase *95% confidence* implies. Can we say that the mean drying time lies in the interval with probability 95%?
- (e) Now construct a 90% confidence interval for the mean drying time by changing one line of the code.
- (f) Many cardiac patients were implanted pacemakers to control their heartbeat. A plastic connector module mounts on the top of the pacemaker. Assuming a standard deviation of 0.0015 and an approximate normal distribution, find a 95% confidence interval for the mean of all connector modules made by a certain company. A random sample of 75 modules has an average of 0.310 inch.

```
>> clear all
>> mx=0.310;
>> s=0.0015;
>> n=75
>> alpha=0.05; % for 95% confidence level
>> zv=norminv(1-alpha/2) % z-value for two-sided CI
>> me=zv*s/sqrt(n); % margin of error
>> CI1= mx-me; % lower CI bound
>> CI2= mx+me; % upper CI bound
```

- (g) Download data from WORKHRS.xls from Blackboard. Find a 95% confidence interval for estimating the mean using both z-score and t-score. Compare the results.
- (h) Download data from ROBOTS.xls from Blackboard. Estimate  $\mu$  of wheels on all social robots with 99% confidence. In repeated sampling, what proportion of confidence intervals will contain the true mean?

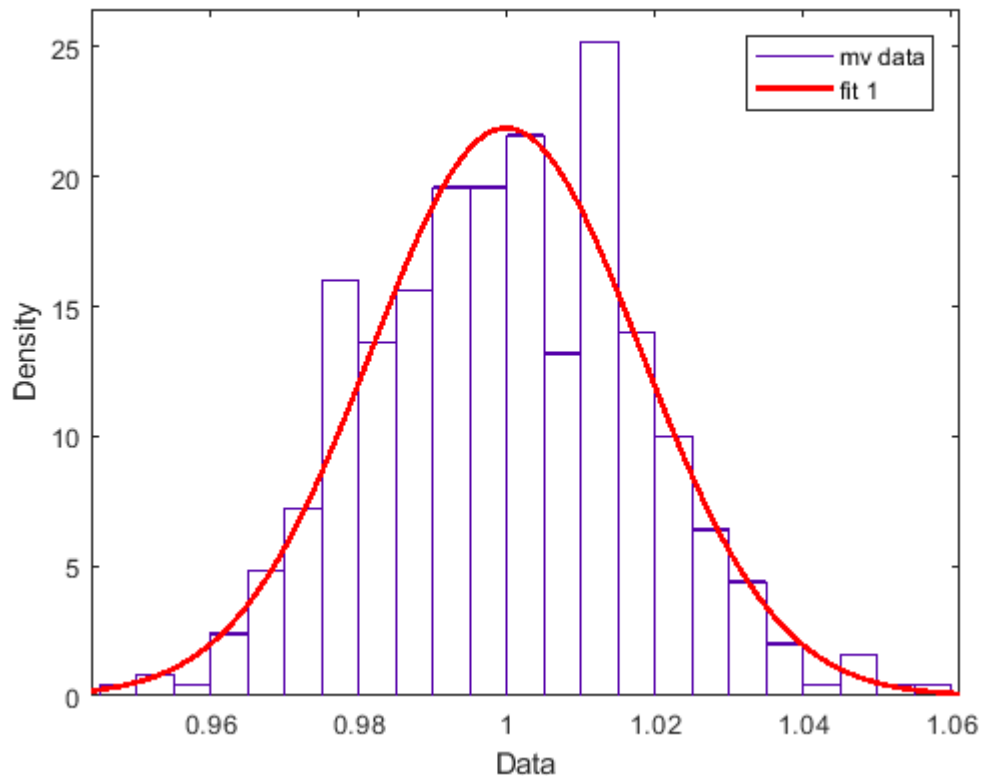
**2 Random sampling from distributions (Monte Carlo simulations):** Use the commands `***rnd` (e.g., `normrnd`, `weibullrnd`, ...). For visual representation of data use `Distribution fitter` in Apps.

- (a) Below is the example of the code that generates normal sample of 100 elements with  $\mu = 1$  and  $\sigma = 0.2$ . The experiment is repeated 500 times. The graph shows the distribution (approximated to normal) of the means from simulation of a random sampling.

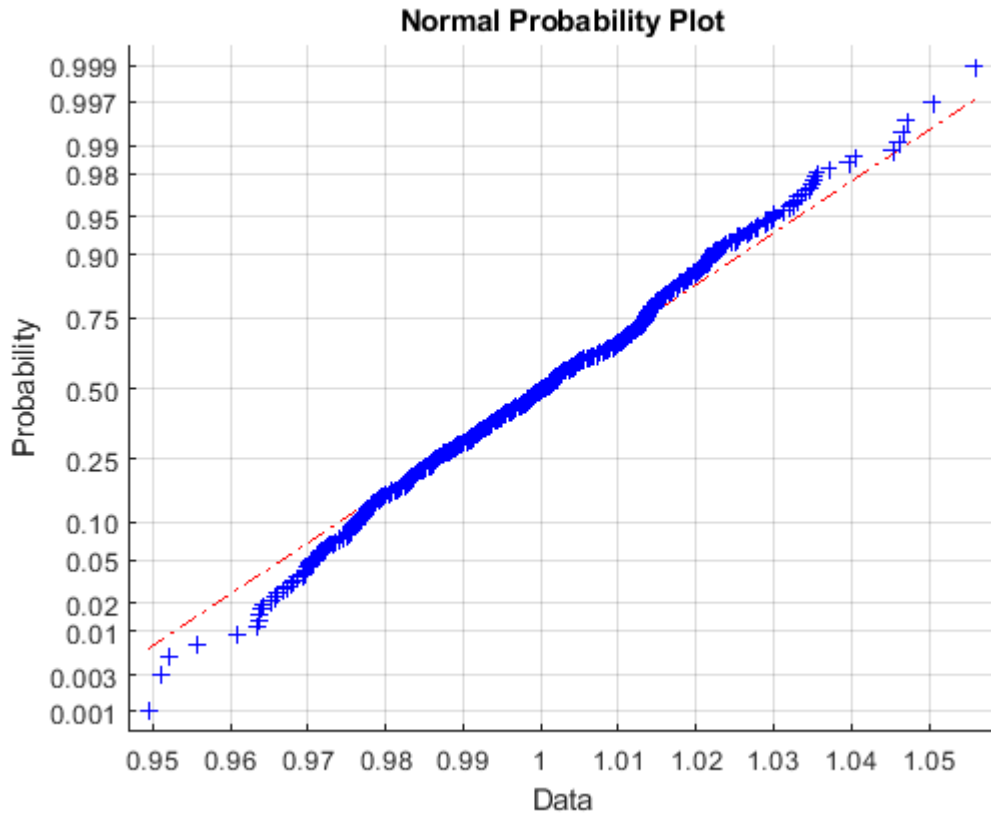
```
>> clear all
>> rng(31) % choose your seed
>> for i = 1 : 500 % take 500 random samples
>> x = normrnd(1,0.2, [1 100]) % each sample contains 100 elements
>> m = mean(x)
>> mv(i) = m; % save means of samples
>> end
```

Mean: 0.999886  
Variance: 0.000332487

Parameter	Estimate	Std. Err.
mu	0.999886	0.000815459
sigma	0.0182342	0.000577484



Use `normplot(mv)` to assess visually whether the sample data comes from a population with a normal distribution (see graph below).



- (b) There is a normal population of rainfalls with mean 900 mm and standard deviation 200 mm. Show using simulations that the more data that is collected, the more accurate the  $\mu$  estimate and hence the smaller its standard error.  
Do you expect any 'unlucky samples' which mean  $\bar{X}$  may not be included into 95% confidence interval?
- (c) Show that the sample mean is less variable than individual values from the population so the mean's standard error is less than the standard deviation.  
*Note that standard deviation describes the variability of sample values; the standard error of an estimator describes the variability of that estimator.*
- (d) Compare mean estimator with a median estimator. Which one has a less spread of errors? Are they all unbiased?

### 3 Central Limit Theorem:

Even if population is not normally distributed, the Central Limit Theorem states that the sampling distribution of the sample mean becomes normal for  $n > 30$ .

- (a) You can run the code below to verify that the distribution of means is approximately normal in Poisson distribution for large  $n$ .

```
clear all
rng(31) % choose your seed
for i = 1 :500
x = poissrnd(2,[1 100])
pm = mean(x)
pmv(i) = pm;
end
```

- (b) Change the code for  $n < 30$ .
- (c) When sample size is 30 or more, we consider the sample size to be large and there is no need to check whether the sample comes from a Normal Distribution. We can use the t-distribution.