

STAT 231

October 5, 2016

Syllabus  $\leq$  Friday, Oct 7.

Tutorial : 3-30  $\rightarrow$  Surya

STP 105.

6-00  $\rightarrow$  Cynthia.

DC

# Roadmap

## MODEL SELECTION METHODS

- Graphical methods

- Relative Frequency

- cdf

- Q-Q - plot

- Run charts

- Numerical methods: (Ch 7)

- Observed versus Expected frequencies

DATA.



MODEL (use Stat. theory)



ESTIMATE

Test the model

and see the

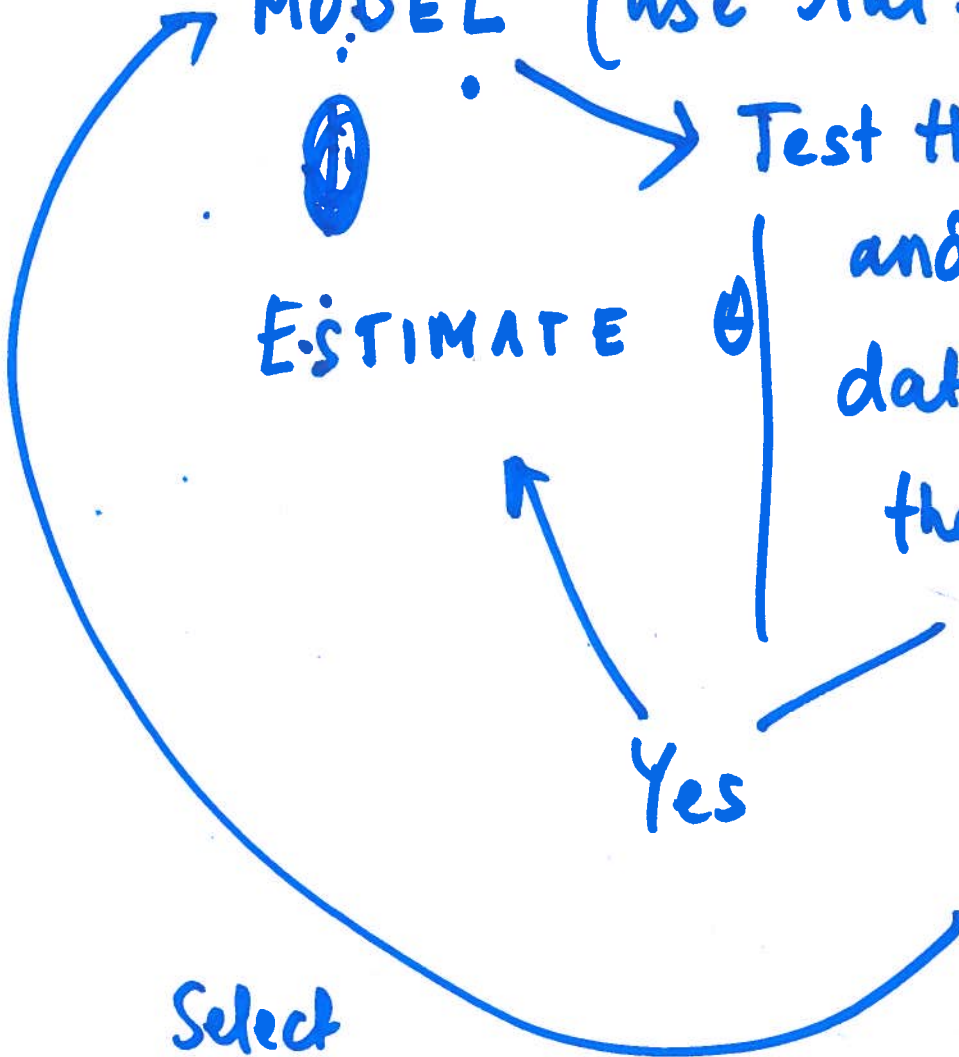
data fits

the model ?

Yes

No

Select  
a different model.



How to select the "right" model?

Graphical ways

- Relative frequency histogram method.

Example: Suppose the model chosen is Gaussian  $(\mu, \sigma)$ .



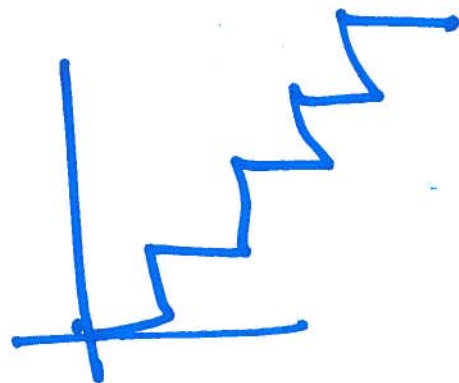
The parameters of the superimposed distribution is chosen by the method of max. Likelihood.

- Compar. the empirical cdf with the theoretical cdf.

$$\{y_1, \dots, y_n\}$$

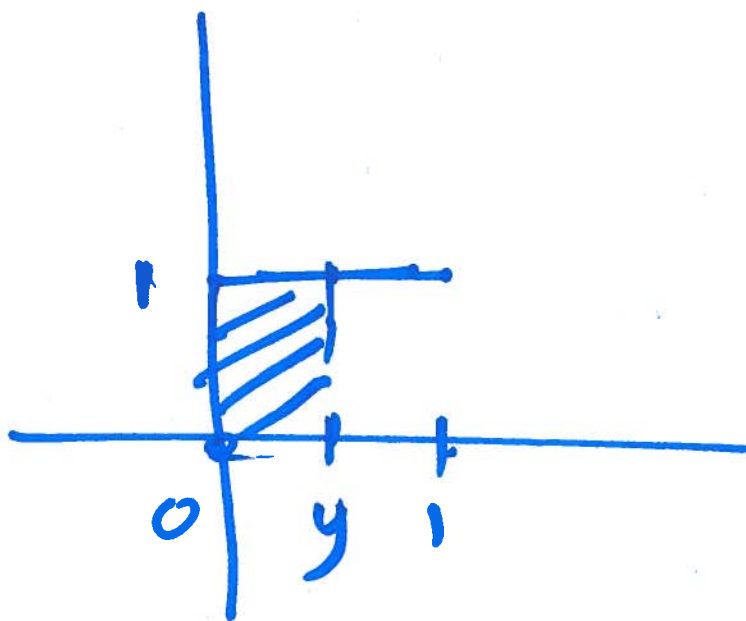
$$\hat{F}(y) = \frac{\# \text{ of obs } \leq y}{n} = \text{ECDF}$$

$$F(y) = P(Y \leq y)$$



Example  $\{y_1, \dots, y_n\}$  DATA

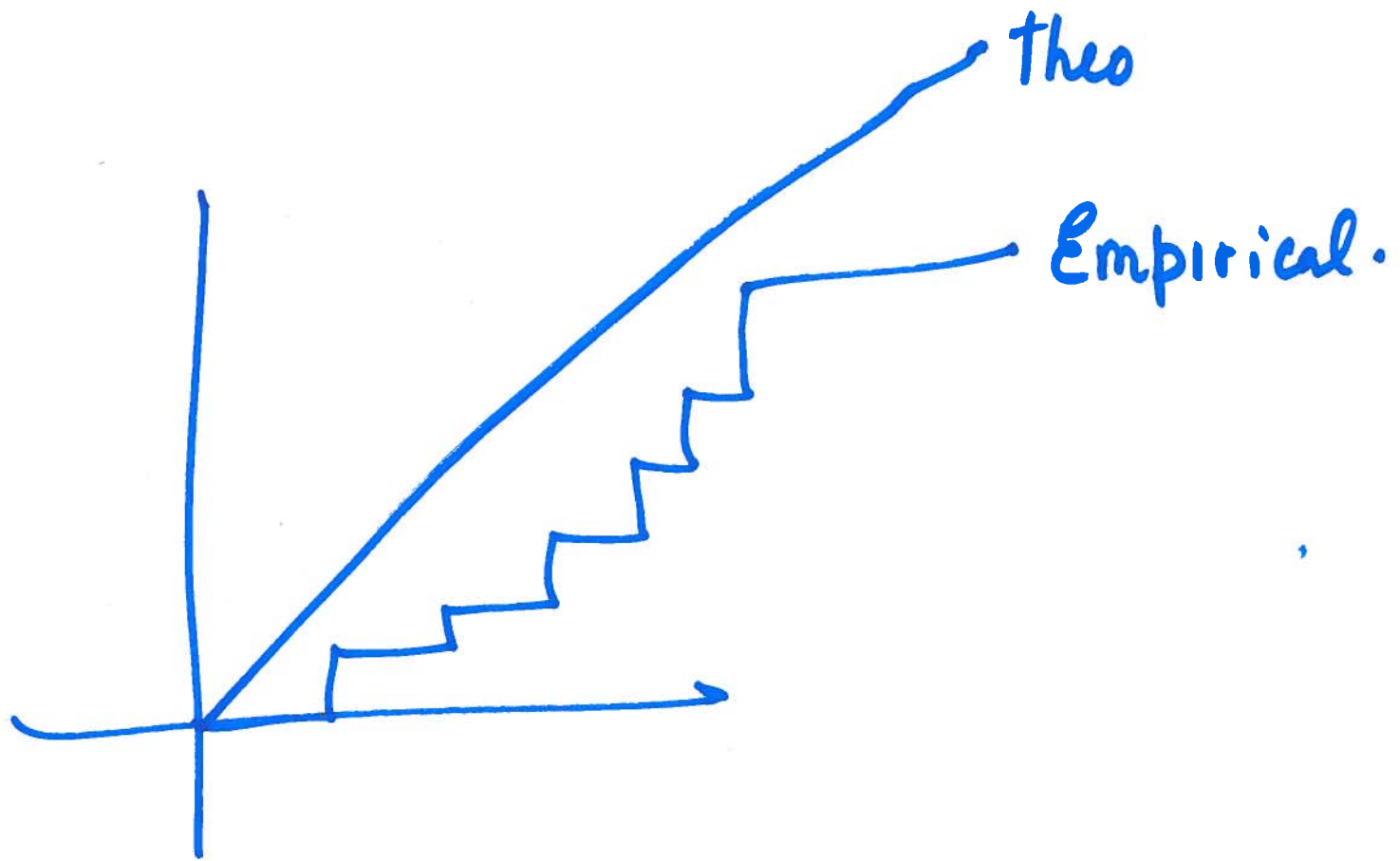
Model:  $Y_i \sim \text{Uni}(0, 1)$   
independent.



$$F(y)$$
$$P(Y \leq y) = y$$

$$\text{cdf of } U(0, 1) = F(y) = y$$
$$= 45^\circ \text{ line}$$





Superimpose the theoretical cdf on the empirical cdf and compare their shapes.

## The Q-Q-plot

Typically, the Q-Q plot is used to check for Gaussian.

Q  $\{y_1, \dots, y_n\}$   
arranged.  $\downarrow$   
 $\{y_{(1)}, y_{(2)}, \dots, y_{(n)}\}$

The Q-Q plot : plots the sample quantiles against the quantiles of the standard normal distribution.

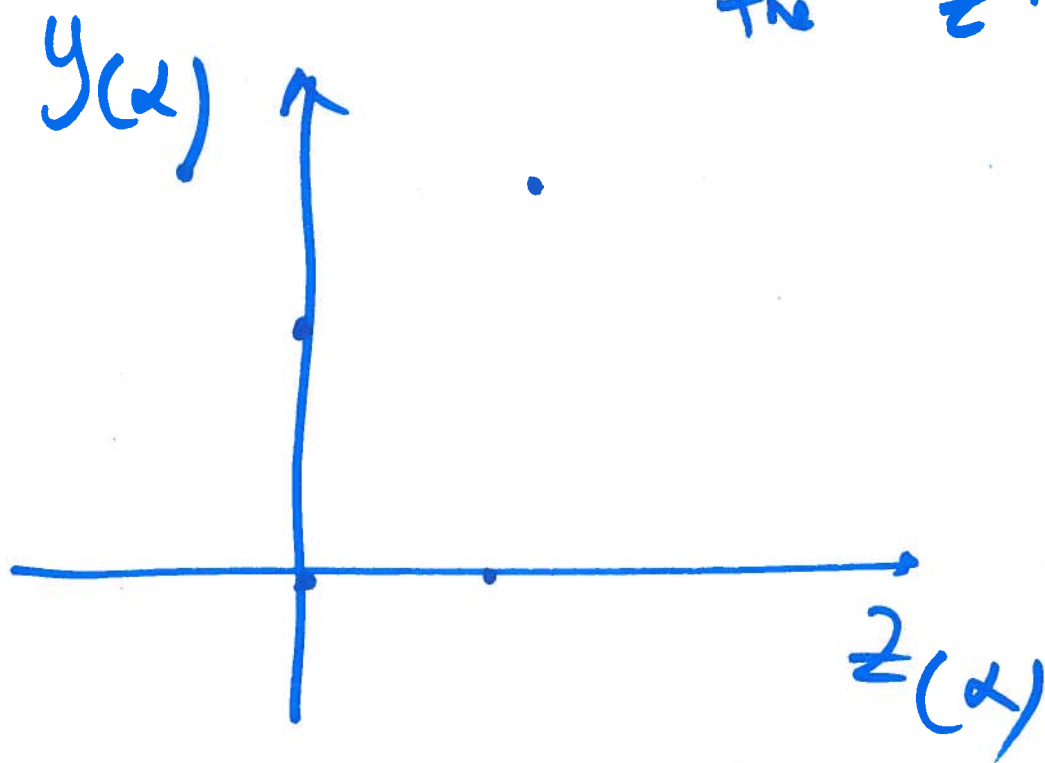
$$Z \sim Q(0, 1)$$



$$Q \rightarrow (y(\alpha), z(\alpha))$$

$y(\alpha) = \alpha^{\text{th}}$  quantile of  
the data set

$z(\alpha) = \alpha^{\text{th}}$  quantile of  
the  $z \sim G(0, 1)$



$y_1, \dots, y_{101}$

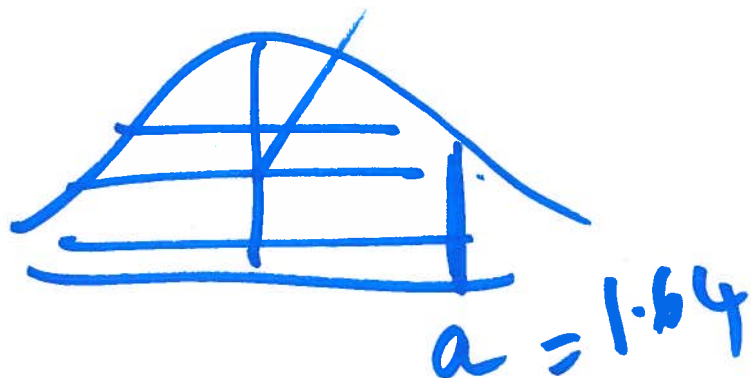
Median  $\rightarrow$   $y$ -value  $\rightarrow$  median of the data

$z$ -value  $\rightarrow$   $Q$

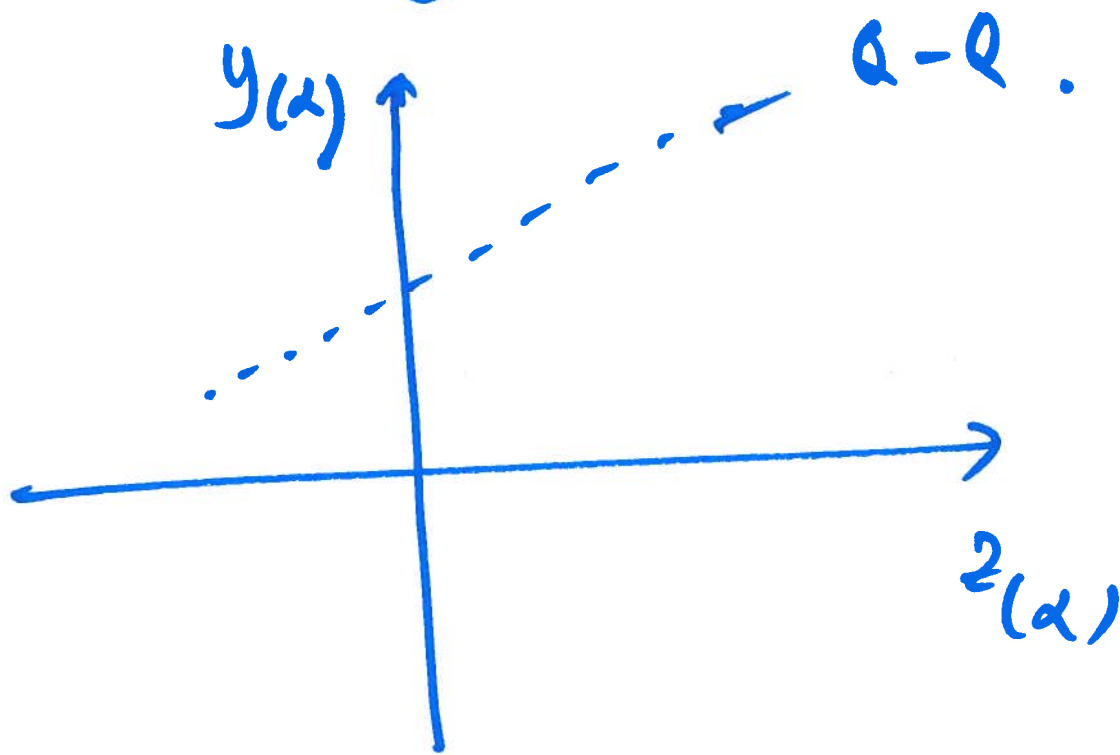
95<sup>th</sup> percentile  $\rightarrow$   $y$ -value  $\rightarrow$  95<sup>th</sup> %ile of the data

0.95

$z$ -value: 95<sup>th</sup> percentile of  $z$

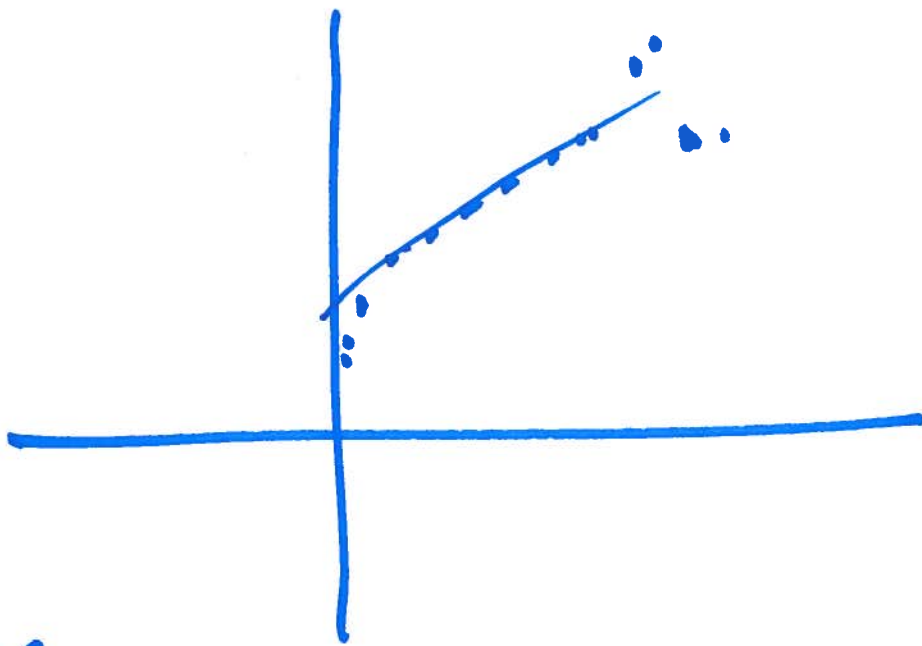


If the  $Q-Q$  plot is a straight line,  
that is strong evidence that the  
data is Gaussian.



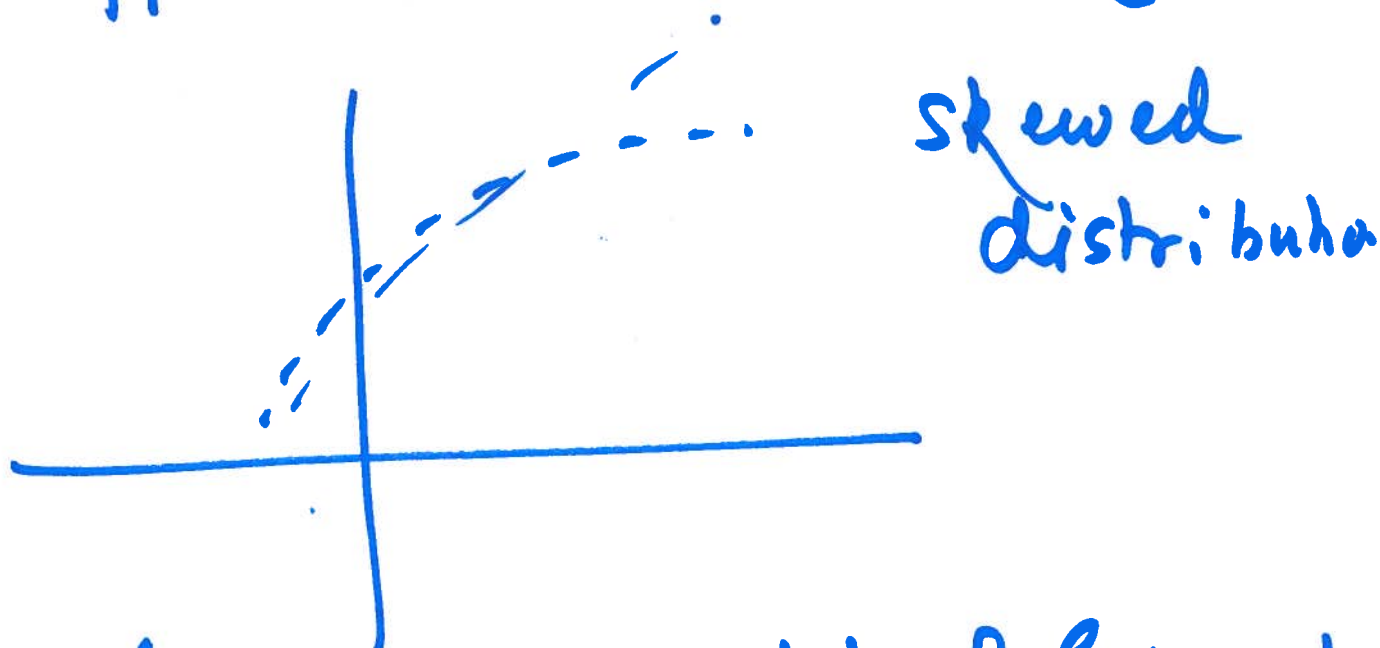
## Notes about the Q-Q plot

- Even if the data is actually Gaussian the extreme points are typically off the line

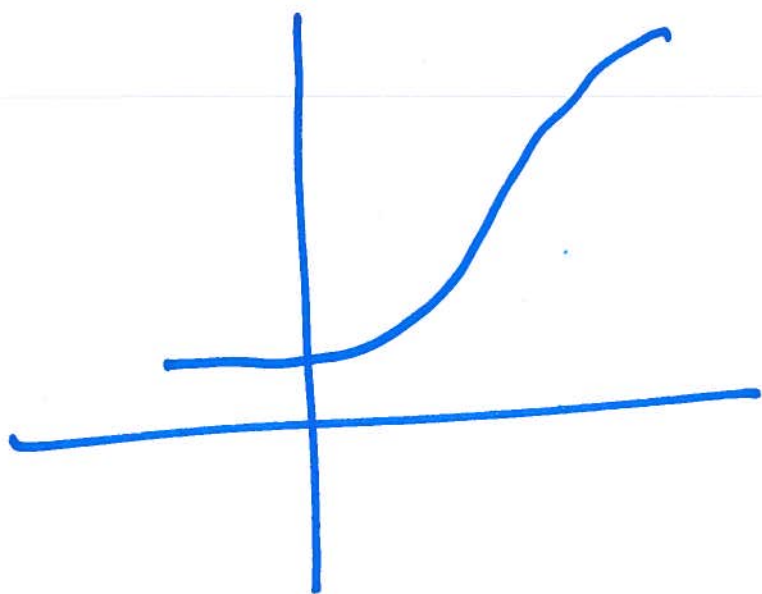


(Why?)

- Apply the Q-Q plot to other distributions. (Exponential)
- Suppose the distribution is asymmetric.



(Try drawing Q-Q plot of Exponential against Normal)



Typically, with a higher kurtosis,  
the Q-Q plot looks like a \$.



# Numerical methods of model checking

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$Y_i$  = # of accidents on a highway in a 1 hour period during rush-hour.

$\{y_1, \dots, y_n\}$

$Y_i \sim \text{Poi}(\mu) \rightarrow \text{SUGGESTED MODEL.}$

# of acci	DATA		Expected frequencies
	Frequency		
0	10		
1	20		
2	40		
3	20		
<del>4</del> 4	10		
5	0		
6	0		
.	1		
.			
.			

Calculate  $\hat{p}$



$$\hat{P}(Y=0) = \frac{e^{-\hat{p}} \hat{p}^0}{0!} = 0\%$$

~~$\times n$~~

Expected frequency  
of 0  $\rightarrow \frac{e^{-\hat{p}} \hat{p}^0}{0!} \times n$

$n$  = sample size

We compare the observed with the expected frequencies and see whether they are "close enough"

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Continuous

$$Y_i \sim \text{Exp}(\lambda)$$
$$y_1, \dots, y_n$$

Data

	Freq
[0, 100]	20
[100, 200]	30
[200, 300]	20
$\approx 300$	10







