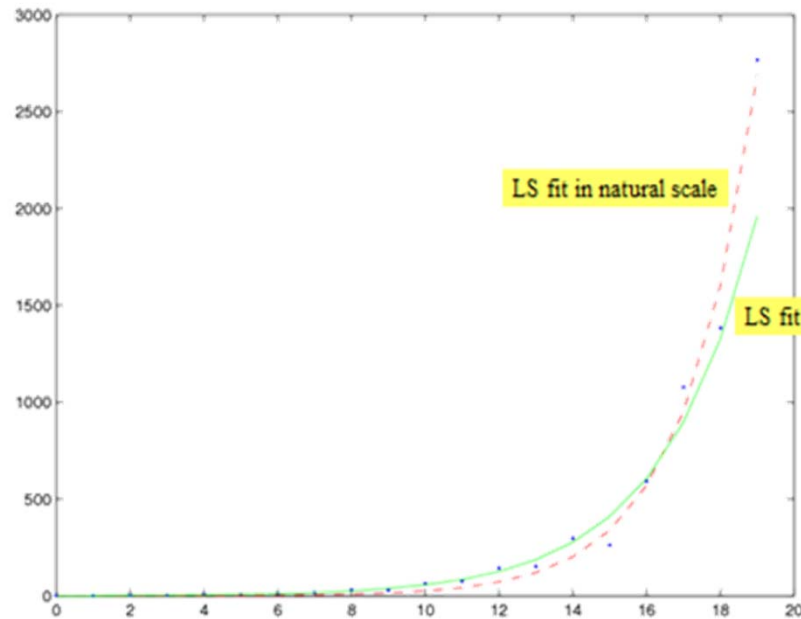
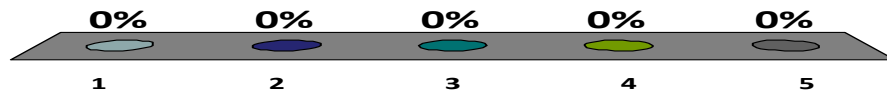


# The residuals for the red estimation are ...

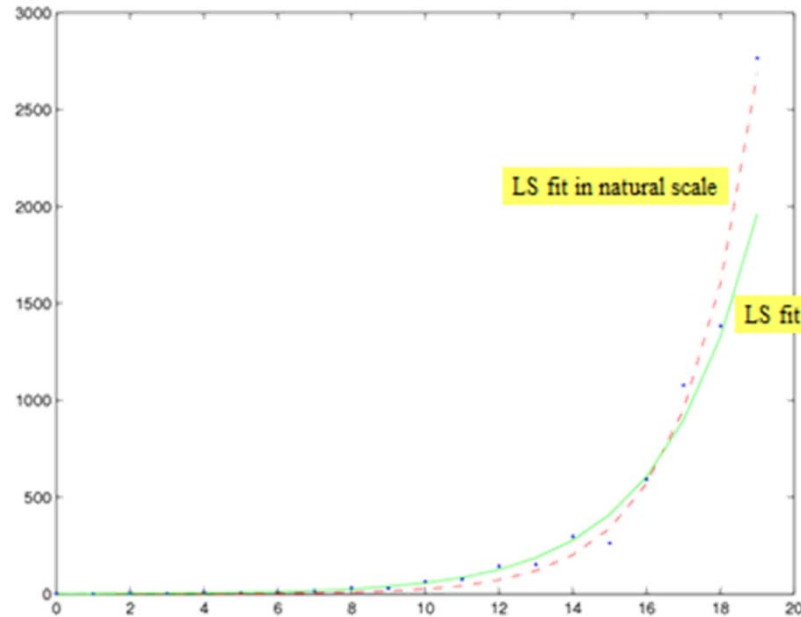


1.  $e_i = y_i - a e^{\alpha t_i}$
2.  $e_i = \frac{y_i - a e^{\alpha t_i}}{a e^{\alpha t_i}}$
3.  $e_i = \log y_i - \log a - \alpha t_i$
4. None of the above
5. I don't know

5

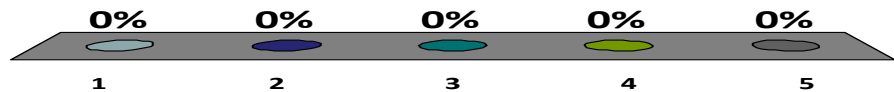


# The residuals for the green estimation are ...



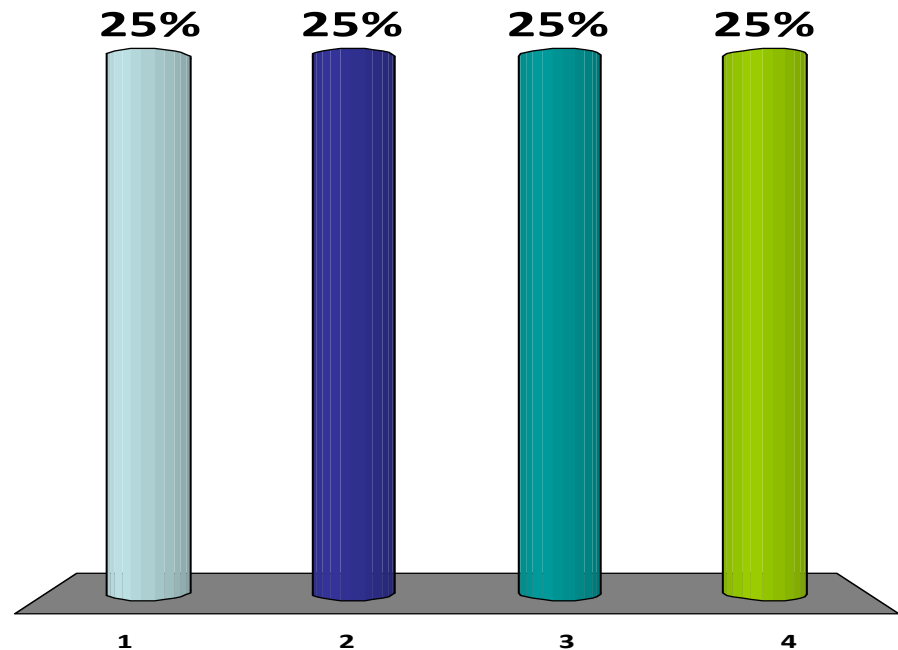
1.  $e_i = y_i - a e^{\alpha t_i}$
2.  $e_i = \frac{y_i - a e^{\alpha t_i}}{a e^{\alpha t_i}}$
3.  $e_i = \log y_i - \log a - \alpha t_i$
4. None of the above
5. I don't know

5



If the error terms  
are not  
homoscedastic, it  
is better to ...

1. Use  $\ell^1$  minimization rather than  $\ell^2$
2. Rescale to make the error term homoscedastic
3. Use  $\ell^2$  minimization rather than  $\ell^1$
4. I do not know

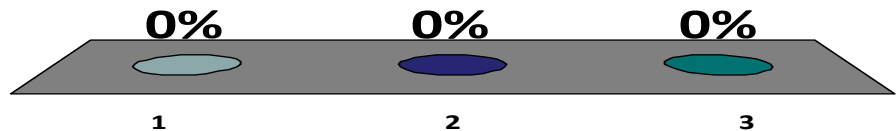
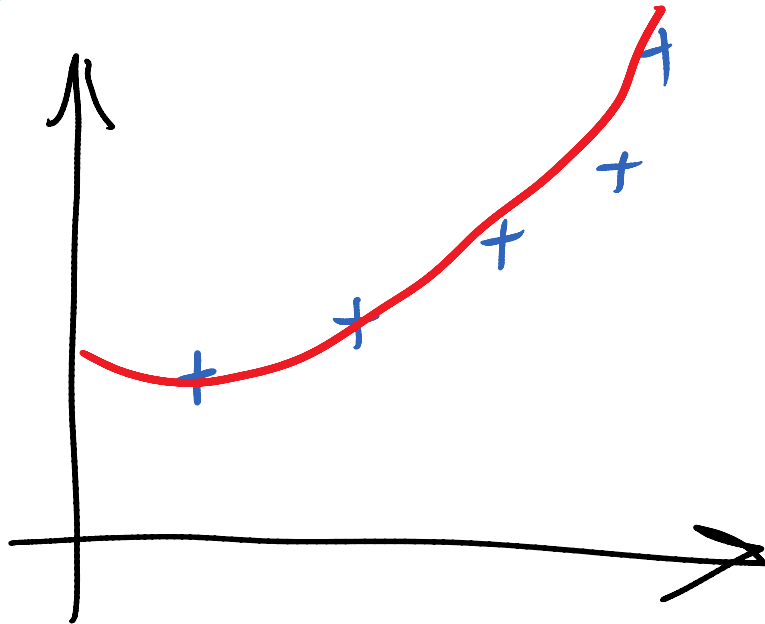


We want to fit a data set  $y_i$  to a polynomial of degree 2:

$$y_i = at_i^2 + bt_i + c.$$

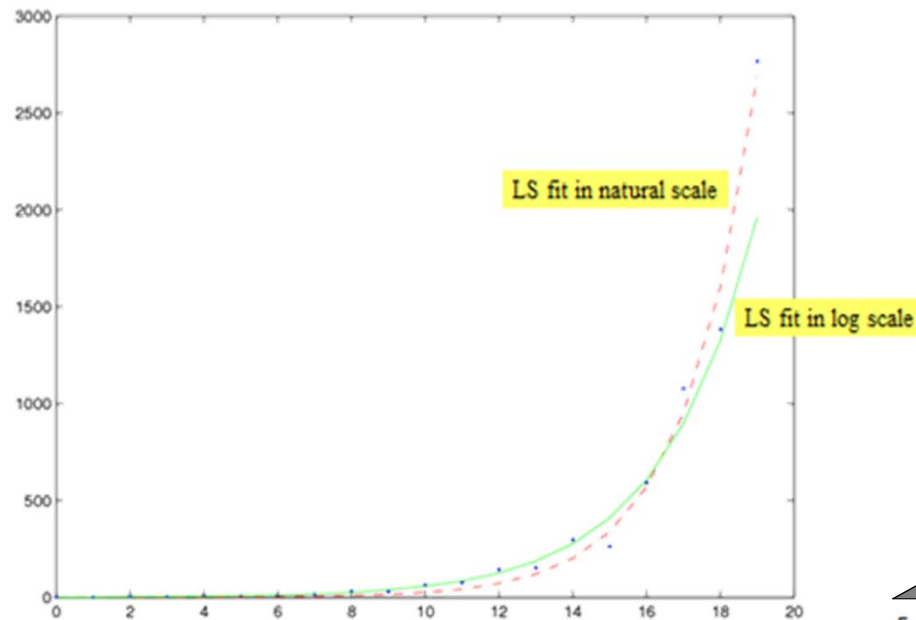
1. Yes
2. No
3. I don't know

Is this a linear regression model ?



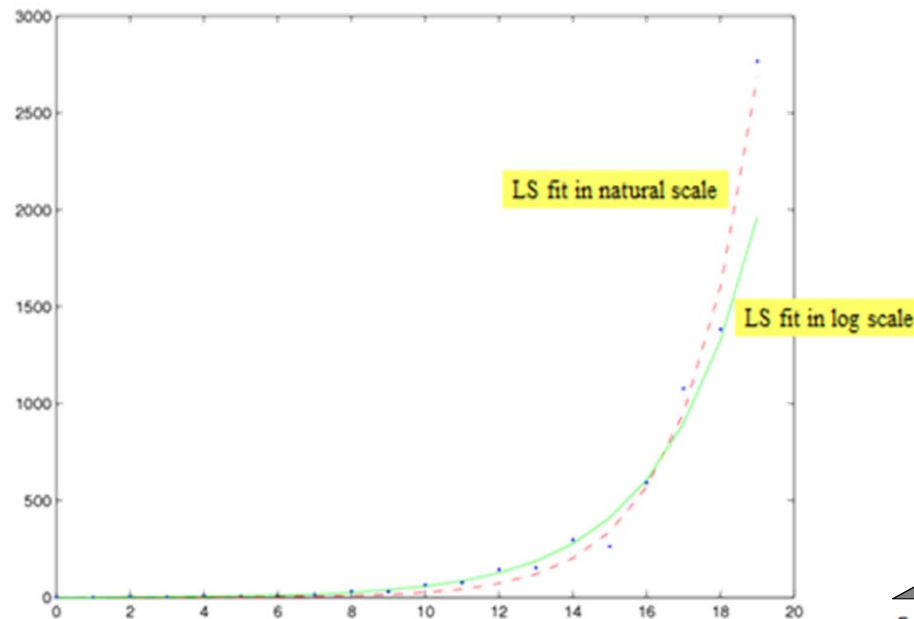
The green estimation corresponds to assuming that the error terms (blue dot – green curve) are iid lognormal.

1. True
2. Almost true
3. False
4. I don't know



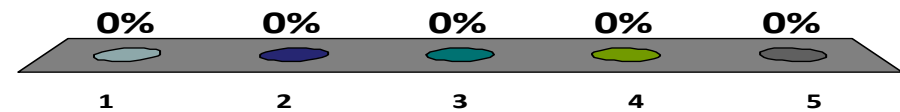
The green estimation corresponds to assuming that the *relative* error terms (blue dot – green curve) are iid normal.

1. True
2. Almost true
3. False
4. I don't know



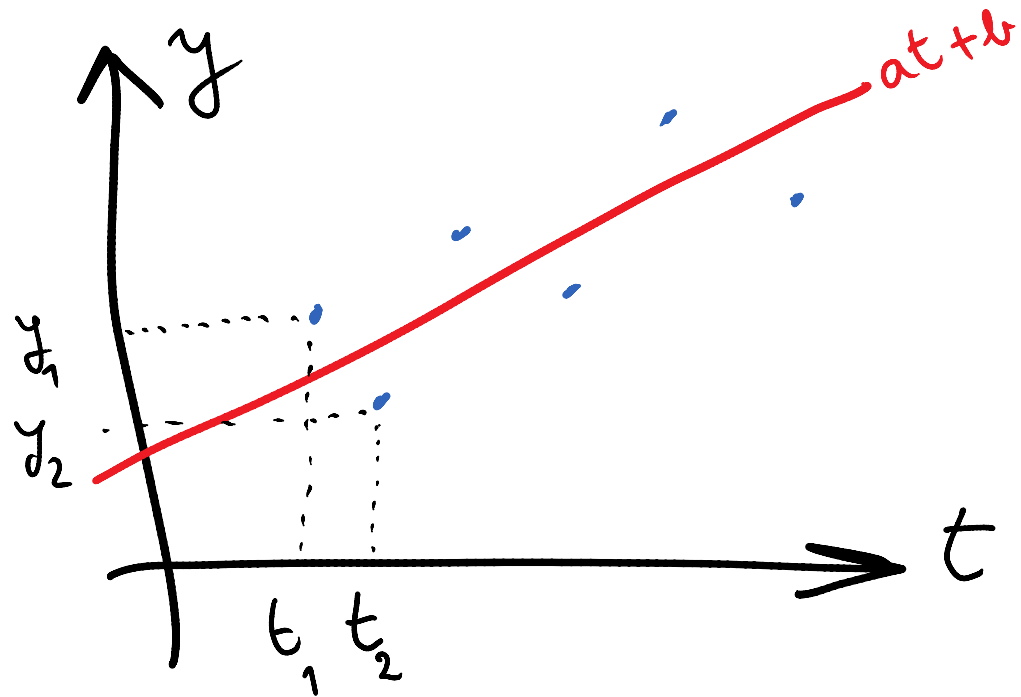
The  $X$  matrix  
for the  
model  $y_i =$   
 $at_i^2 + bt_i + c$   
has full rank.

1. Yes
2. No
3. It depends on the  $t_i$ 's
4. It depends on the  $y_i$ 's
5. I don't know



We fit the model  
 $y_i = at_i + b$   
using least  
squares.

The obtained line  
is such that the  
average distance  
from the points to  
the line is 0.

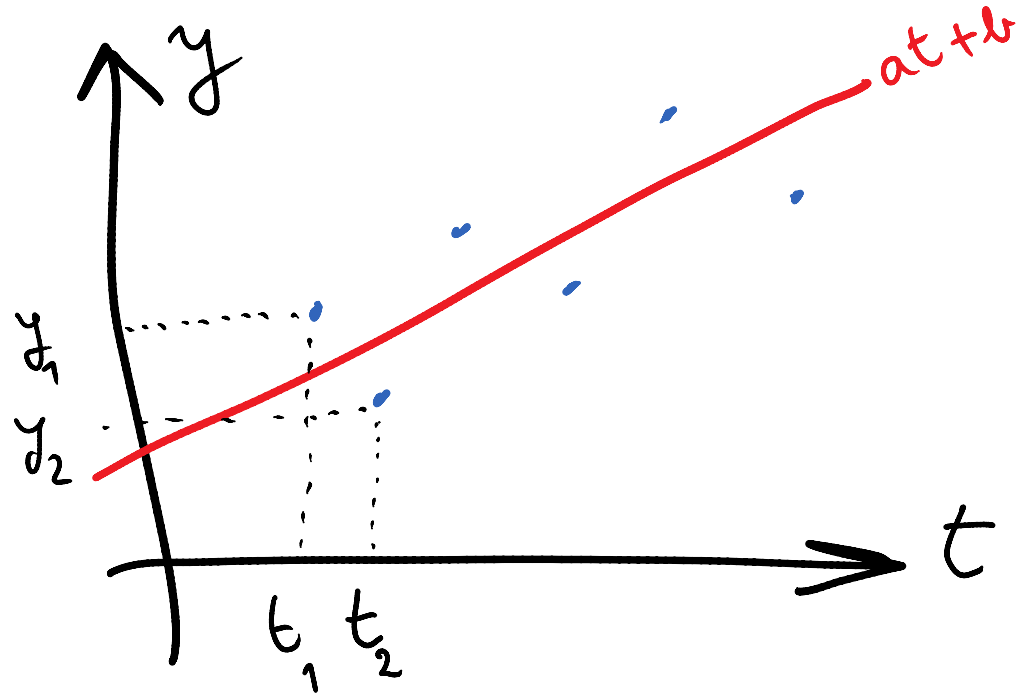


1. True
2. False
3. It depends on the data
4. I don't know



We fit the model  $y_i = at_i + b$  using  $\ell^1$  norm minimization.

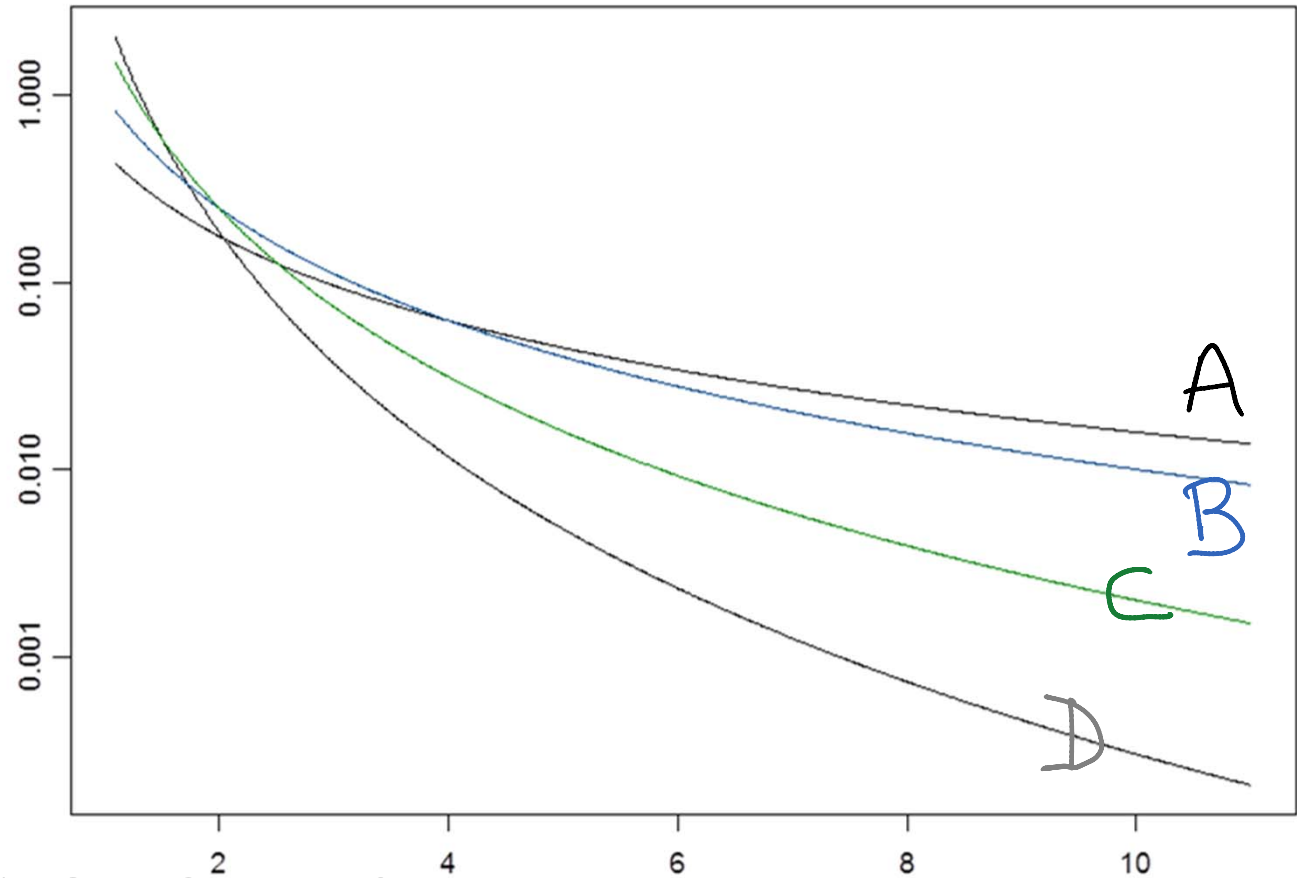
The obtained line leaves an equal number of points on each side



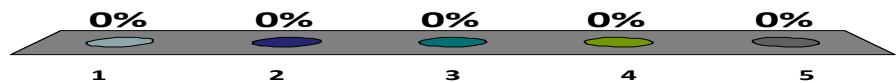
1. True
2. False
3. It depends on the data
4. I don't know



Find the parameter  $p$  for each of these standard Pareto PDFs

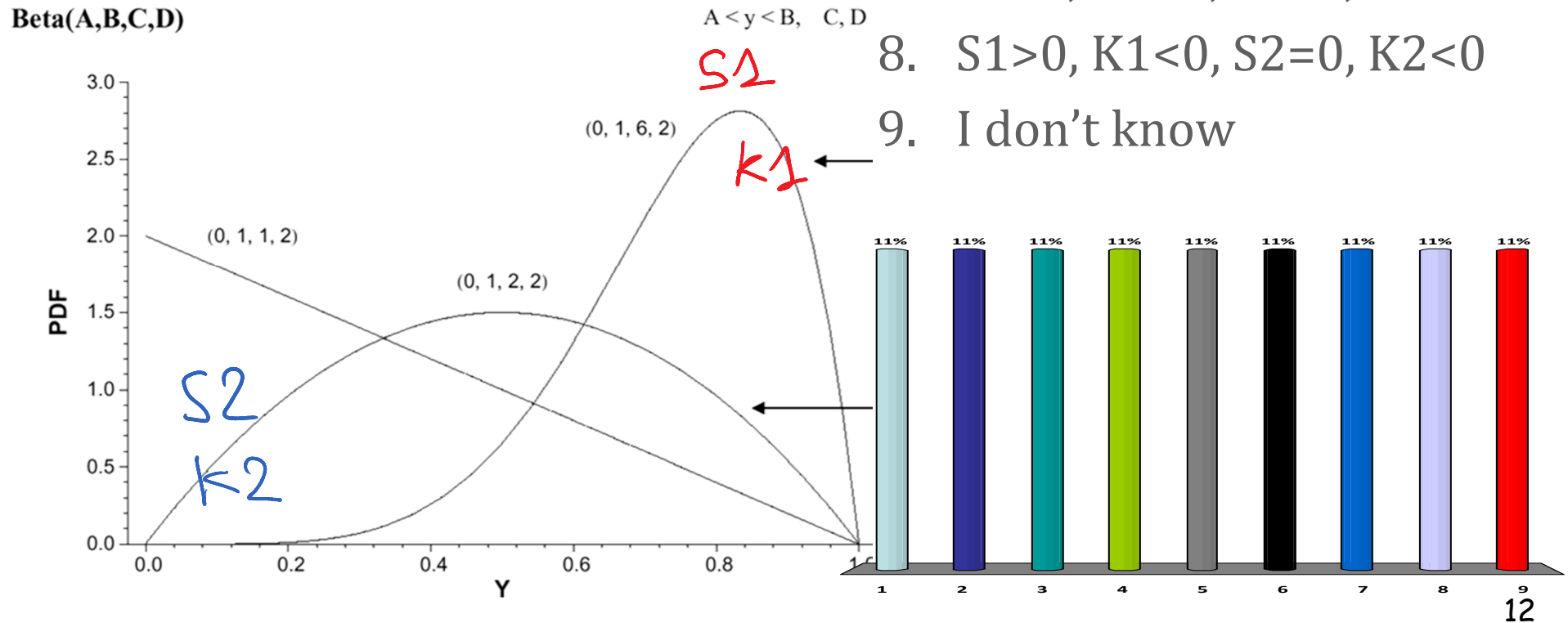


1.  $A = 0.5; B = 1; C = 2; D = 3$
2.  $A = 3; B = 2; C = 1; D = 0.5$
3.  $A = 0.5; B = 2; C = 1; D = 3$
4.  $A = 1; B = 2; C = 3; D = 0.5$
5. I don't know



# The skewness indices $S_1$ , $S_2$ and the Kurtosis indices $K_1$ , $K_2$ are...

1.  $S_1 > 0, K_1 > 0, S_2 > 0, K_2 > 0$
2.  $S_1 < 0, K_1 > 0, S_2 > 0, K_2 < 0$
3.  $S_1 > 0, K_1 > 0, S_2 = 0, K_2 > 0$
4.  $S_1 < 0, K_1 > 0, S_2 = 0, K_2 < 0$
5.  $S_1 > 0, K_1 < 0, S_2 > 0, K_2 > 0$
6.  $S_1 < 0, K_1 < 0, S_2 > 0, K_2 < 0$
7.  $S_1 > 0, K_1 < 0, S_2 = 0, K_2 > 0$
8.  $S_1 > 0, K_1 < 0, S_2 = 0, K_2 < 0$
9. I don't know



If a positive random variable has a finite mean and is heavy tailed, its variance is infinite

1. True
2. False
3. It depends
4. I don't know



The CCDF of a  
Pareto  
distribution  
follows a power  
law...

1. True
2. False
3. It depends
4. I don't know



A Pareto distribution is heavy tailed ...

1. True
2. False
3. It depends
4. I don't know



For a Pareto distribution, the hazard rate  $\lambda(t)$  is such that

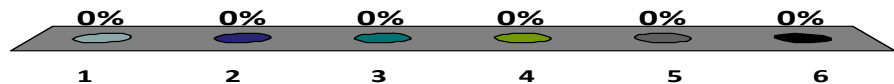
$$\lim_{t \rightarrow \infty} \lambda(t) = 0$$

1. True
2. False
3. It depends
4. I don't know



The distribution of the sum of  $n$  iid random variables with heavy tail and index  $p < 2$ , for large  $n$ , is approximately...

1. Normal
2. Stable with same index  $p$
3. Stable but not necessarily with same index  $p$
4. Poisson
5. It depends
6. I don't know





The distribution of the sum of  $n$  iid random variables with finite variance, for large  $n$ , is approximately...

1. Normal
2. Stable
3. Poisson
4. It depends
5. I don't know

