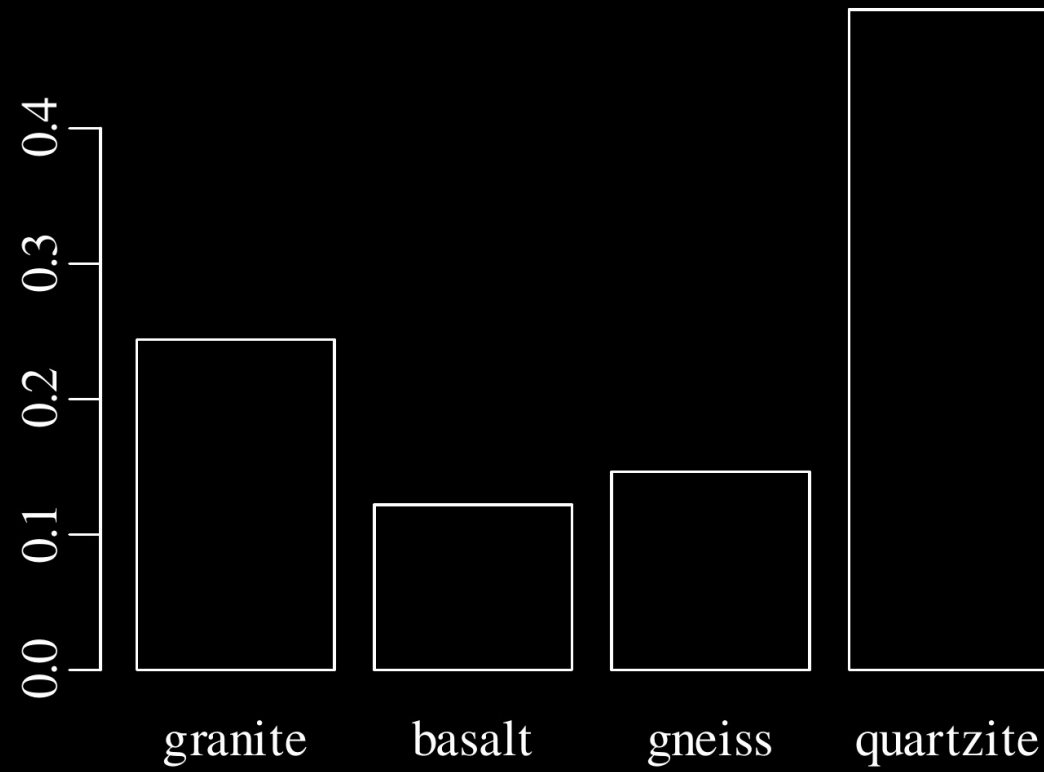


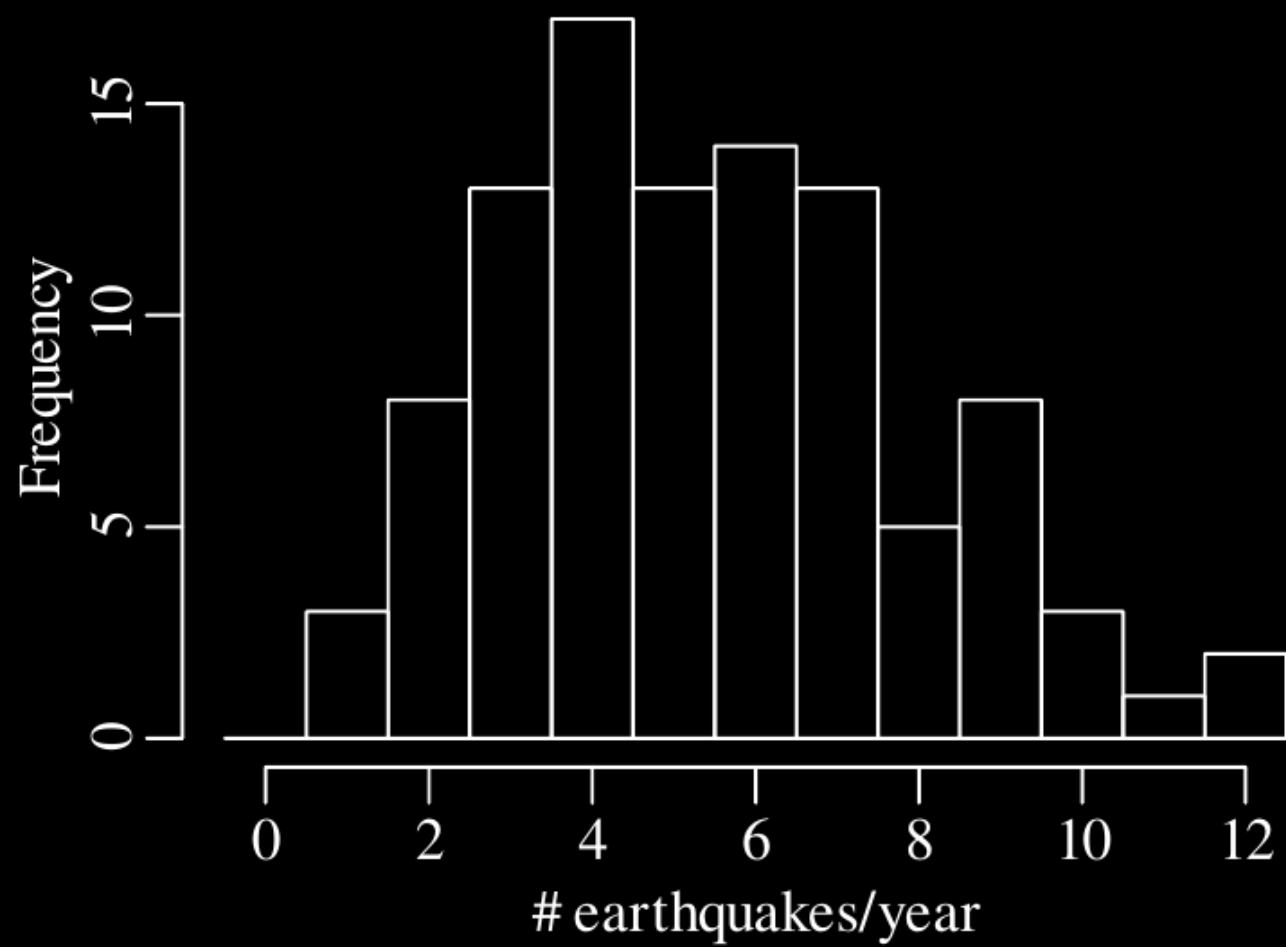
1. What is so ‘normal’ about the Gaussian distribution?

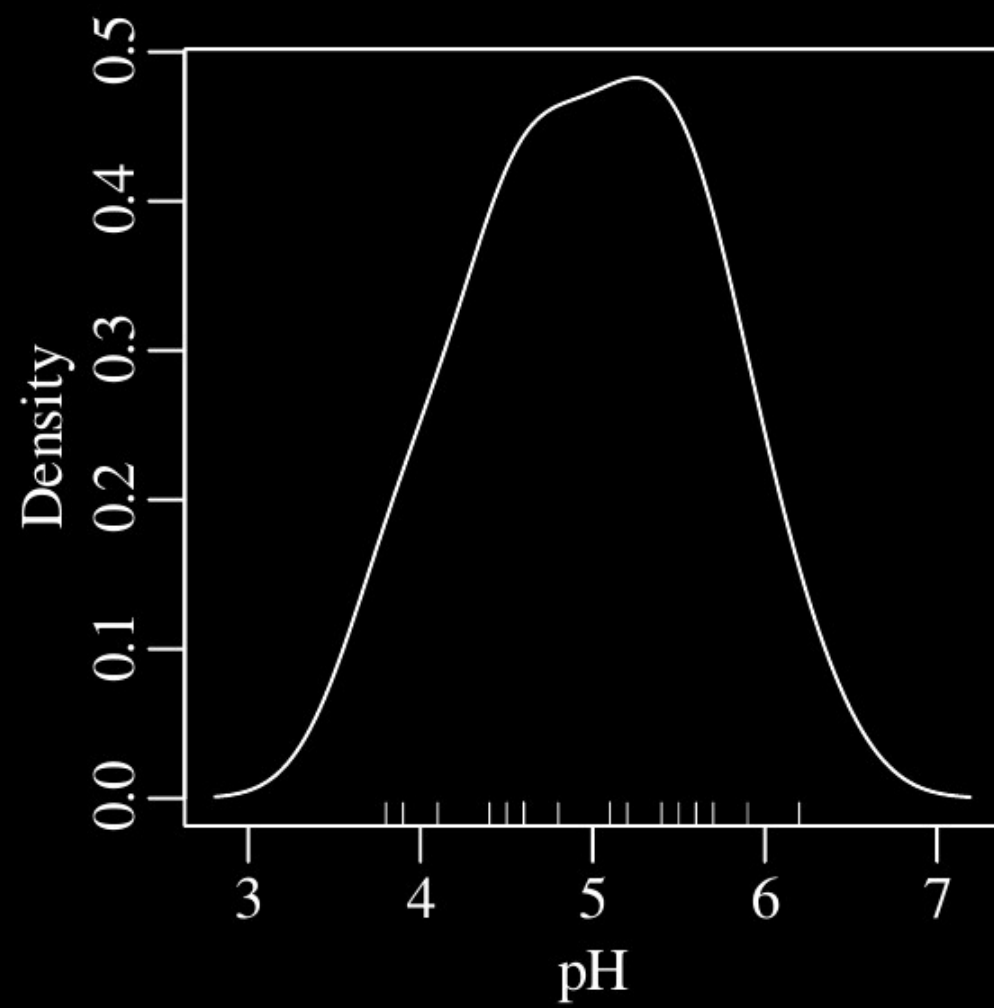
2. Is geology ‘normal’?

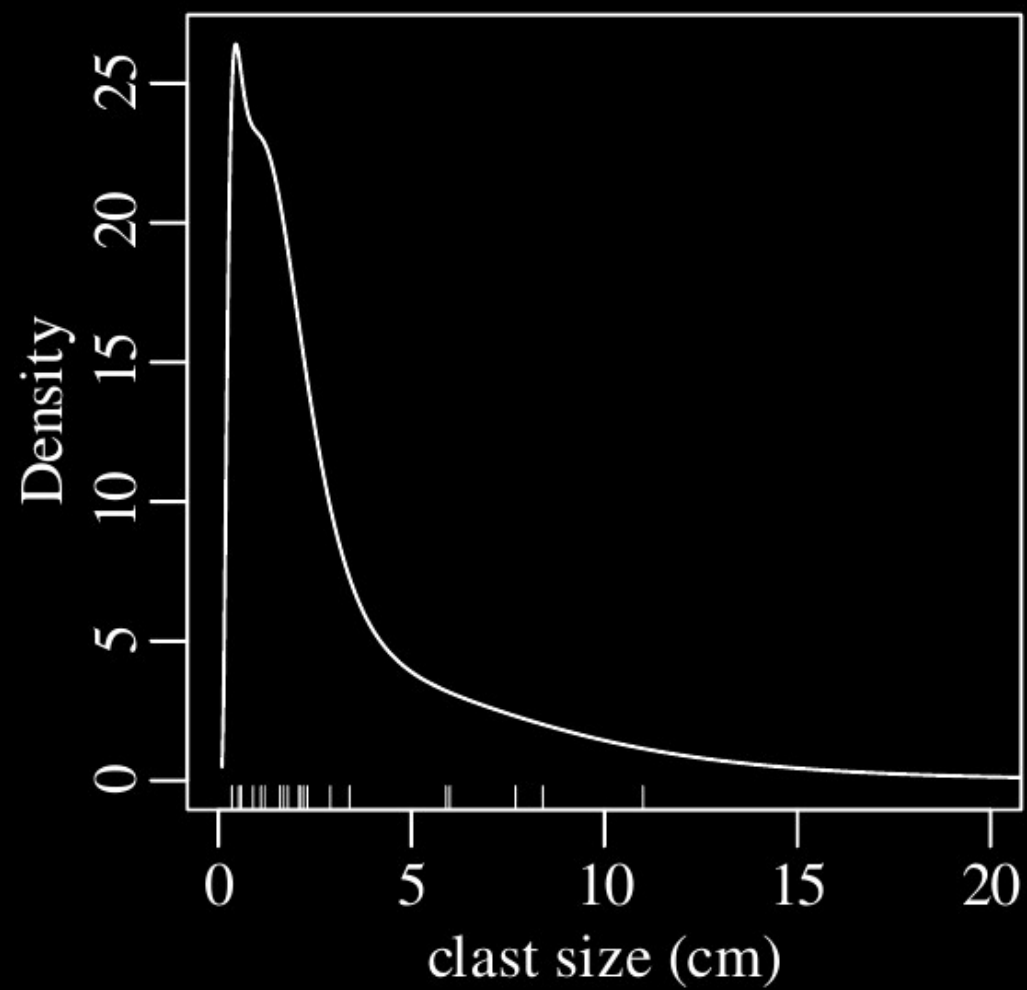
Pieter Vermeesch

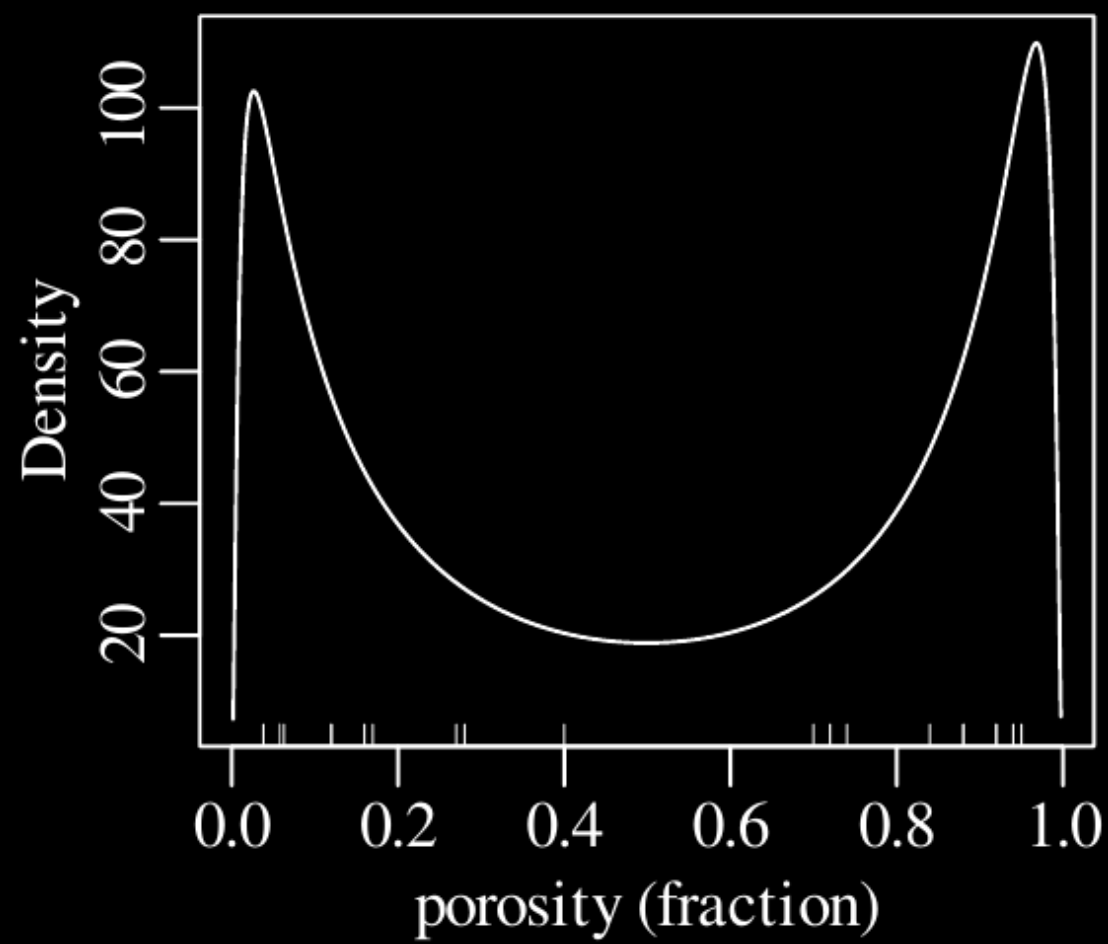
p.vermeesch@ucl.ac.uk

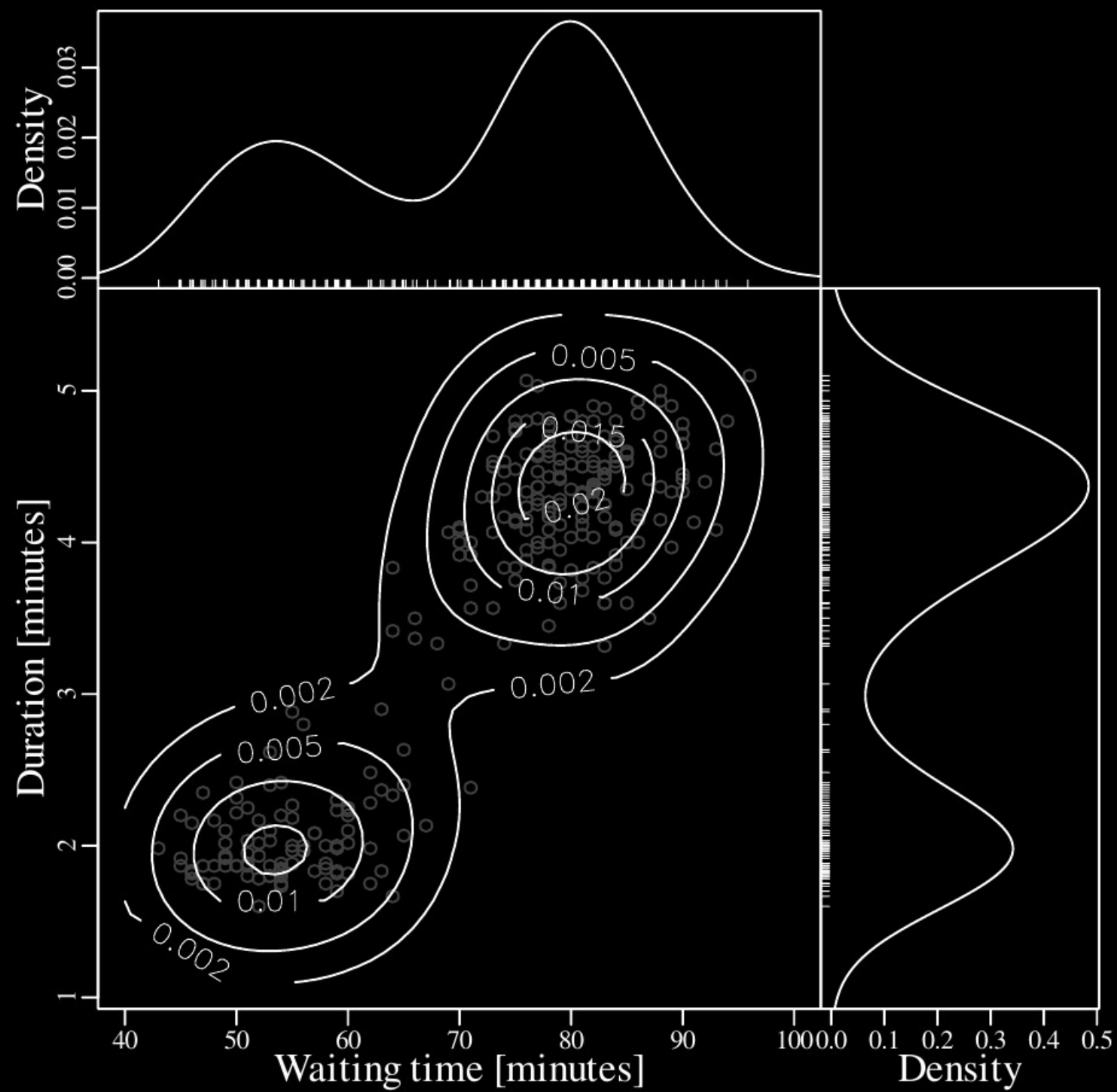


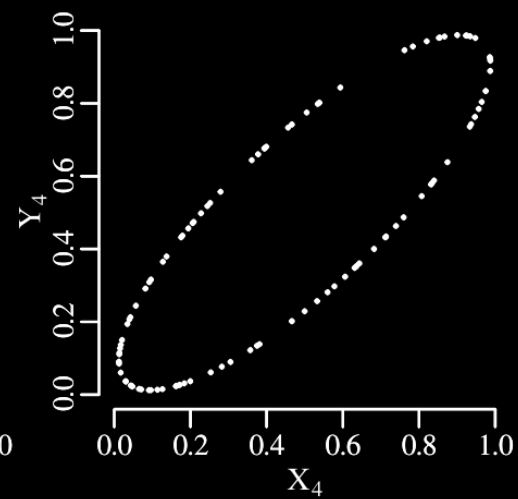
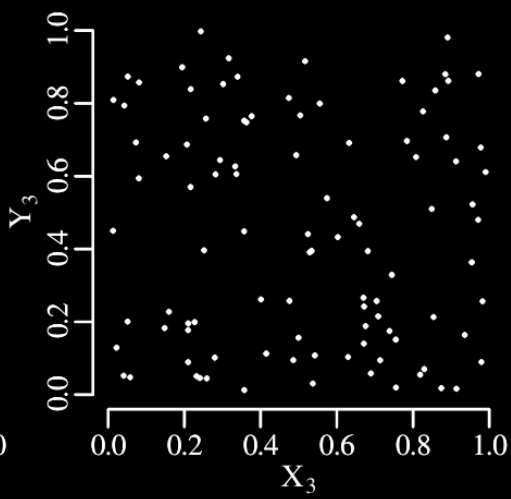
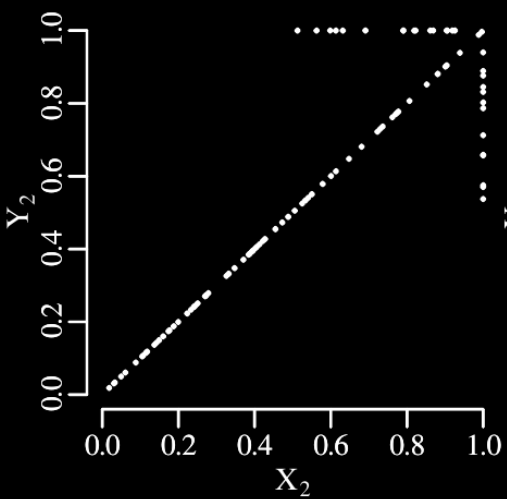
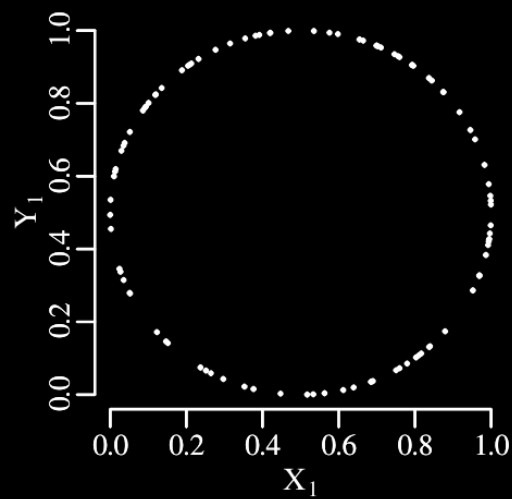
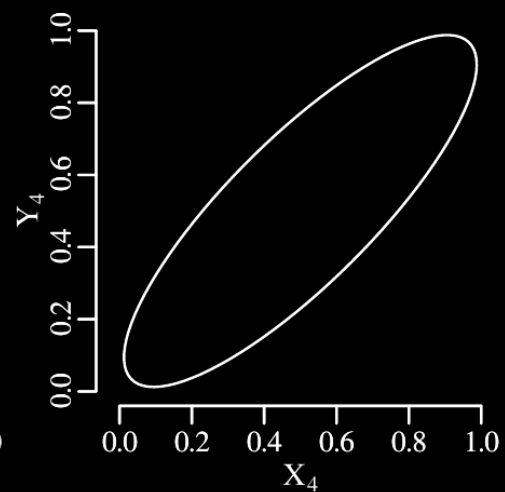
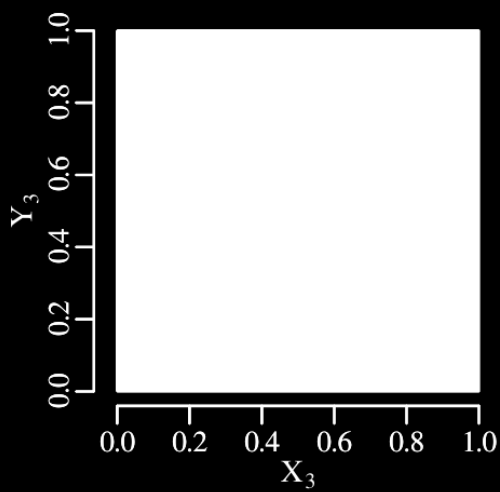
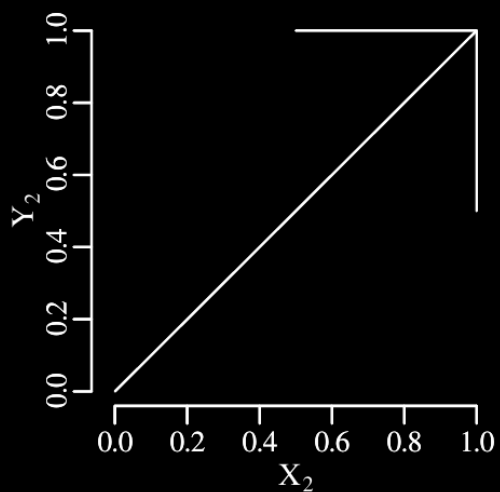
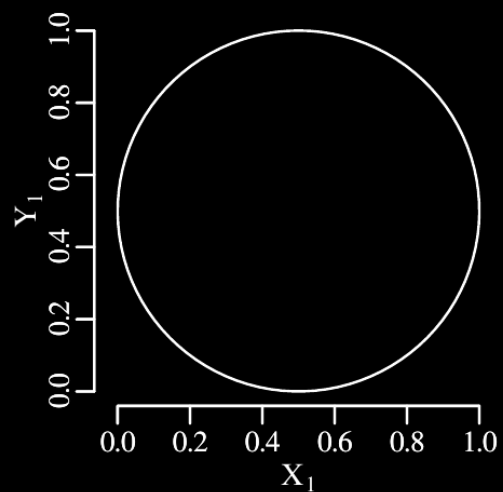






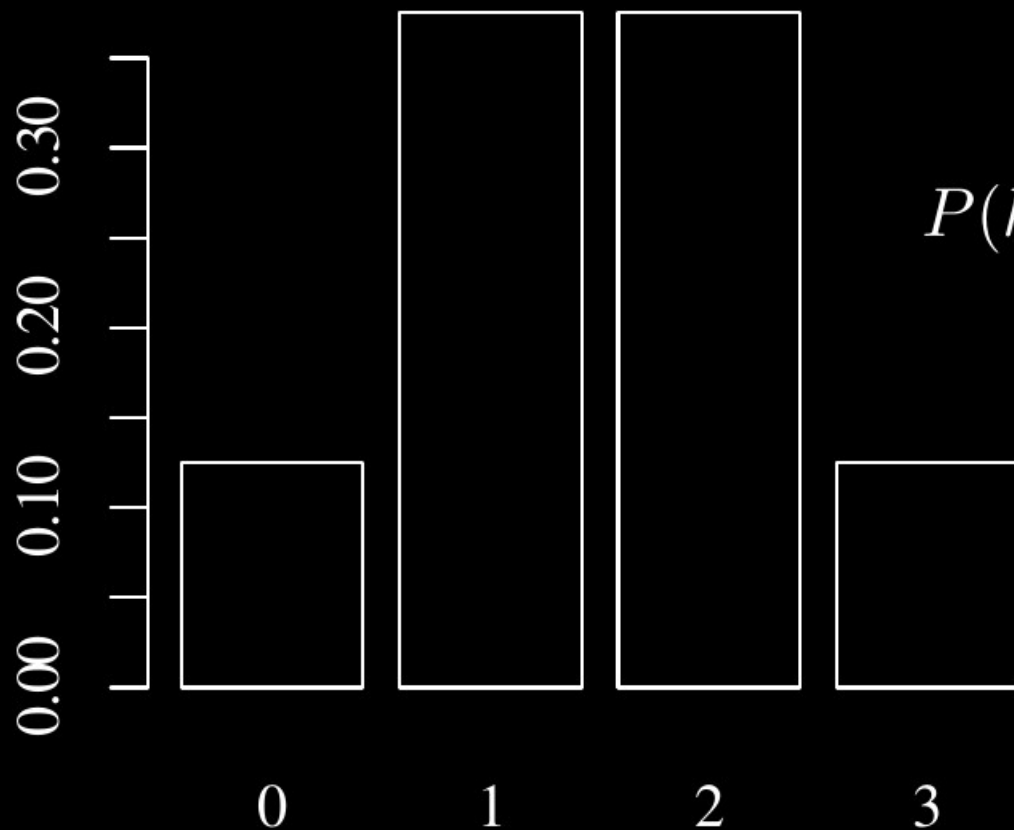






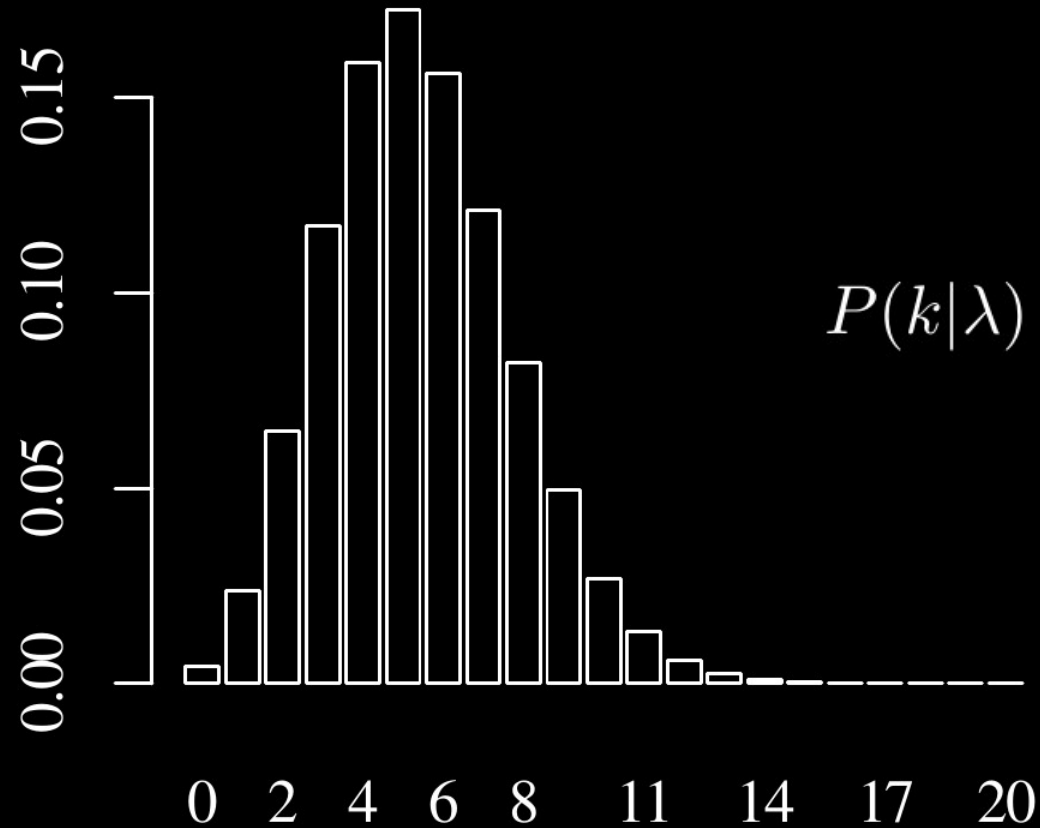
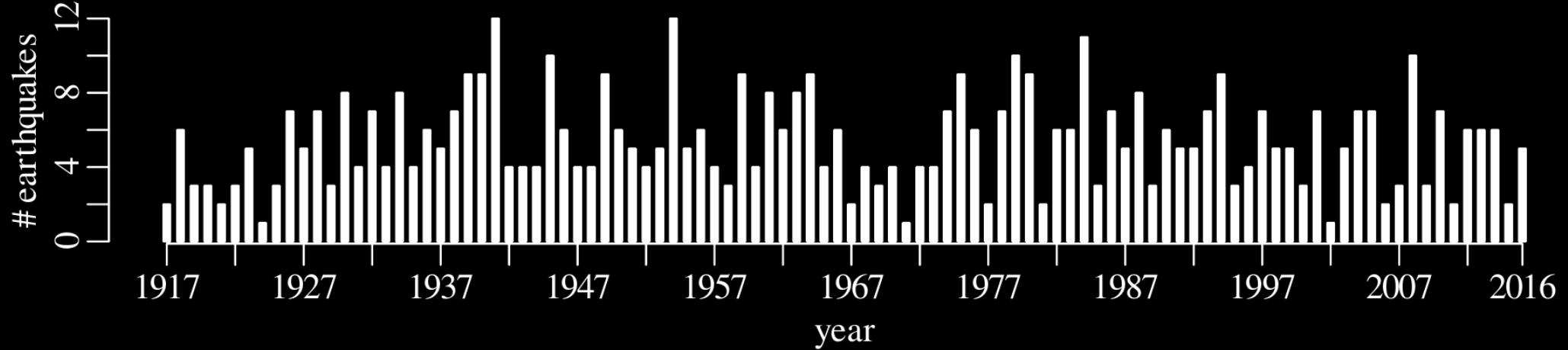
The binomial distribution

$$P(2 \times H \cap 1 \times T) = \frac{\{THH\}\{HTH\}\{HHT\}}{\{HHH\}\{THH\}\{HTH\}\{HHT\}\{TTH\}\{THT\}\{HTT\}\{TTT\}} = \frac{3}{8}$$



$$P(k|n, p) = \binom{n}{k} p^k (1 - p)^{n-k}$$

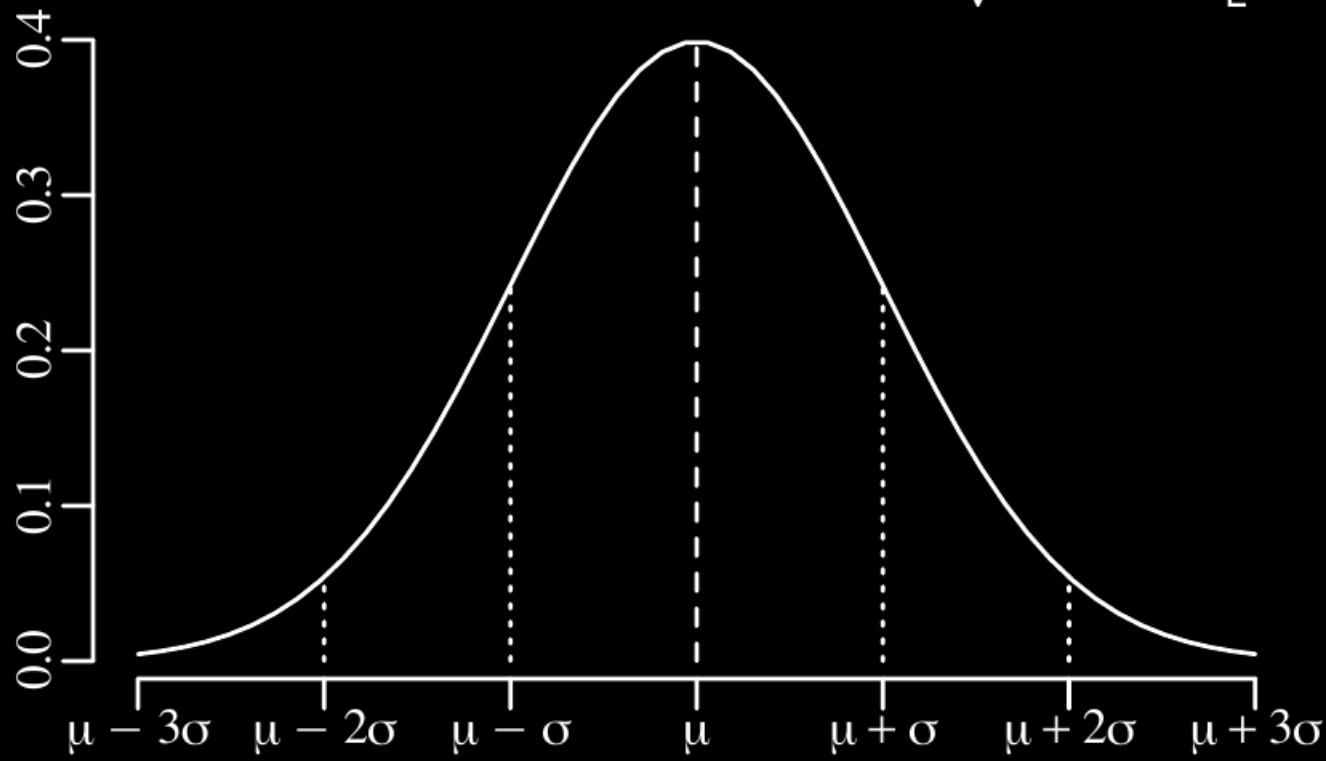
The Poisson distribution



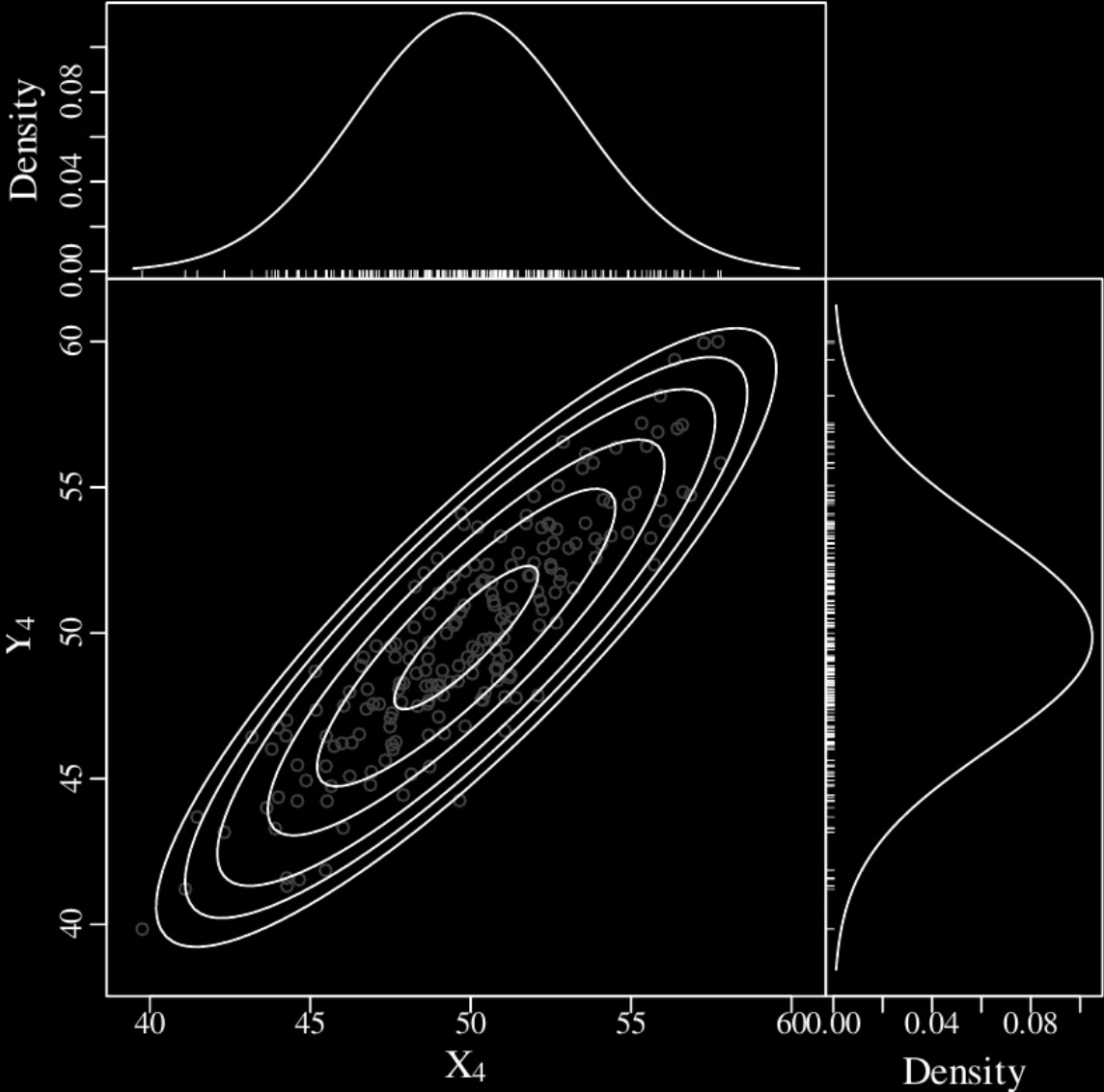
$$P(k|\lambda) = \frac{\lambda^k e^{-\lambda}}{k!}$$

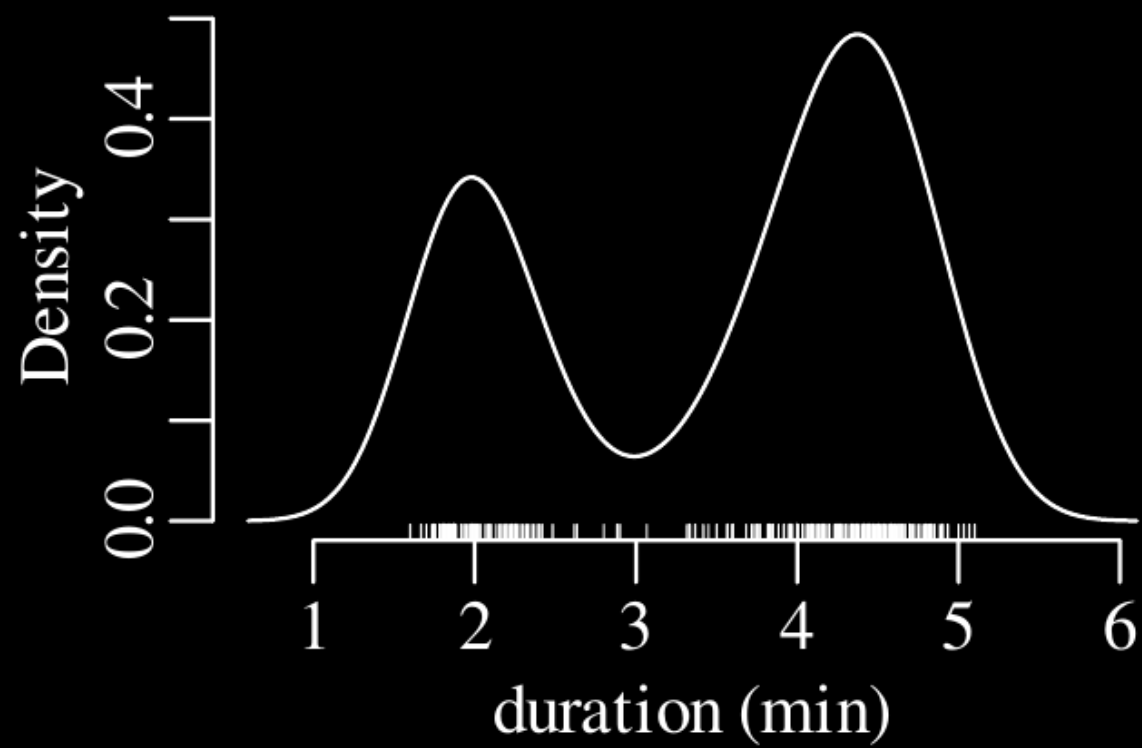
the Gaussian distribution
the **normal** distribution

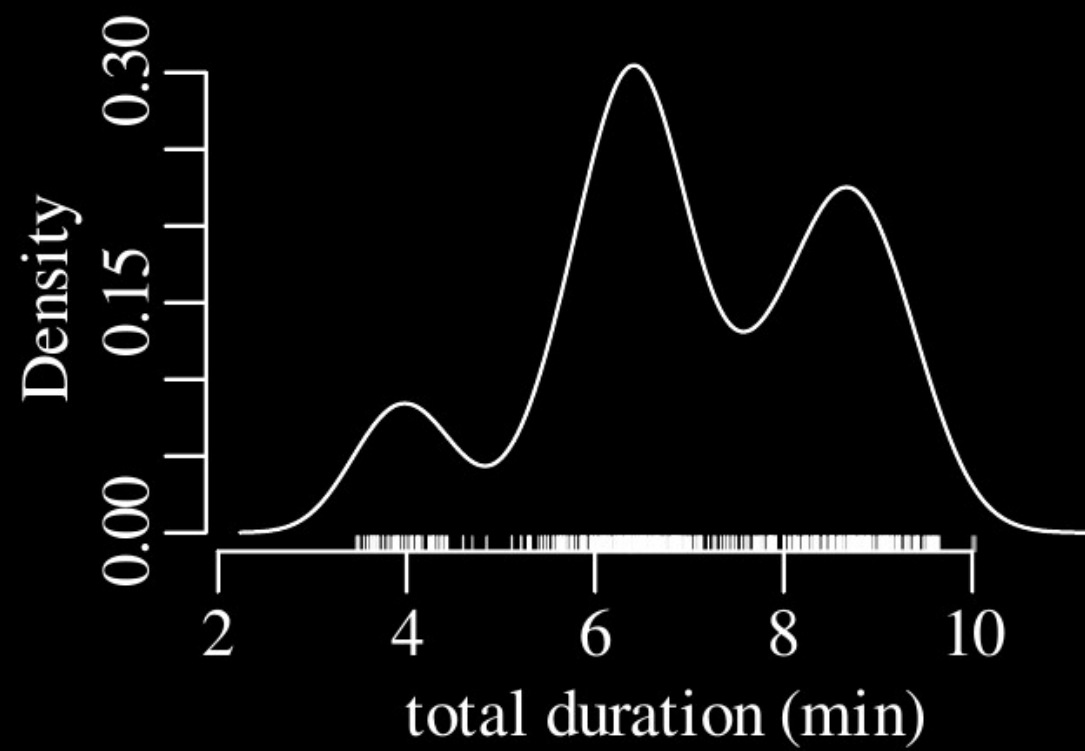
$$f(x|\mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(x - \mu)^2}{2\sigma^2}\right]$$

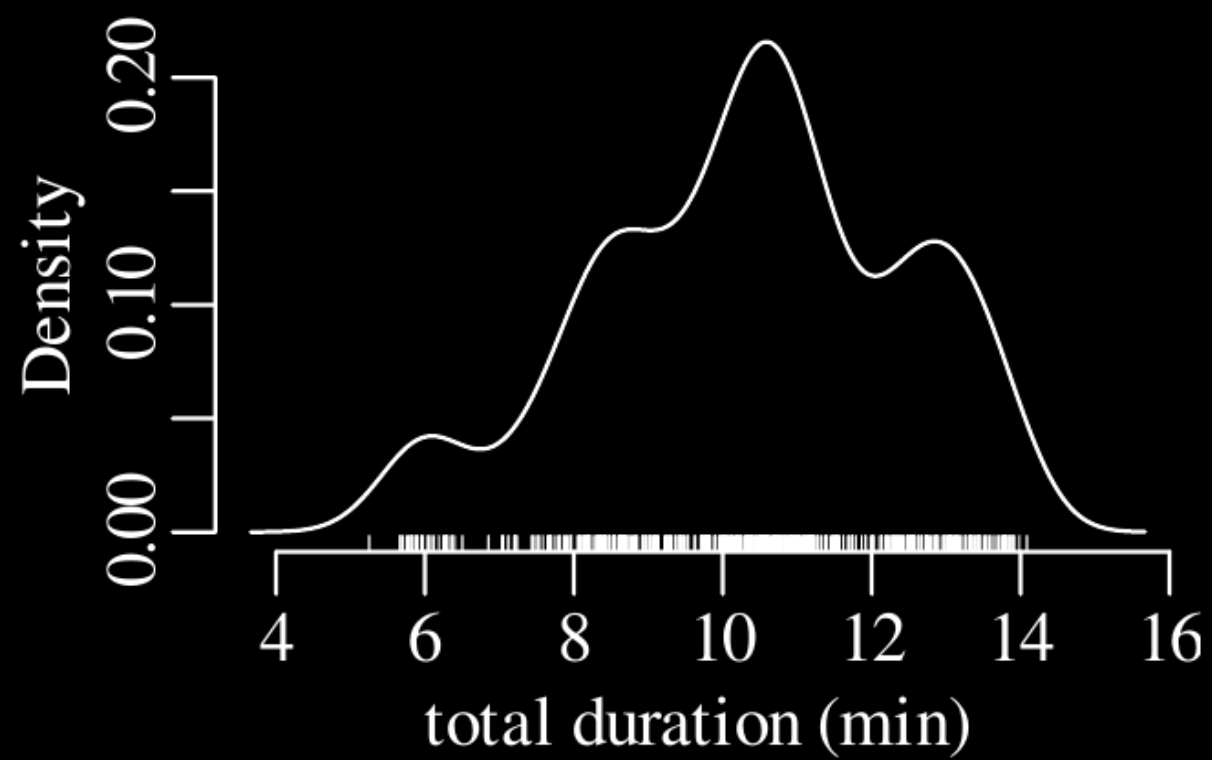


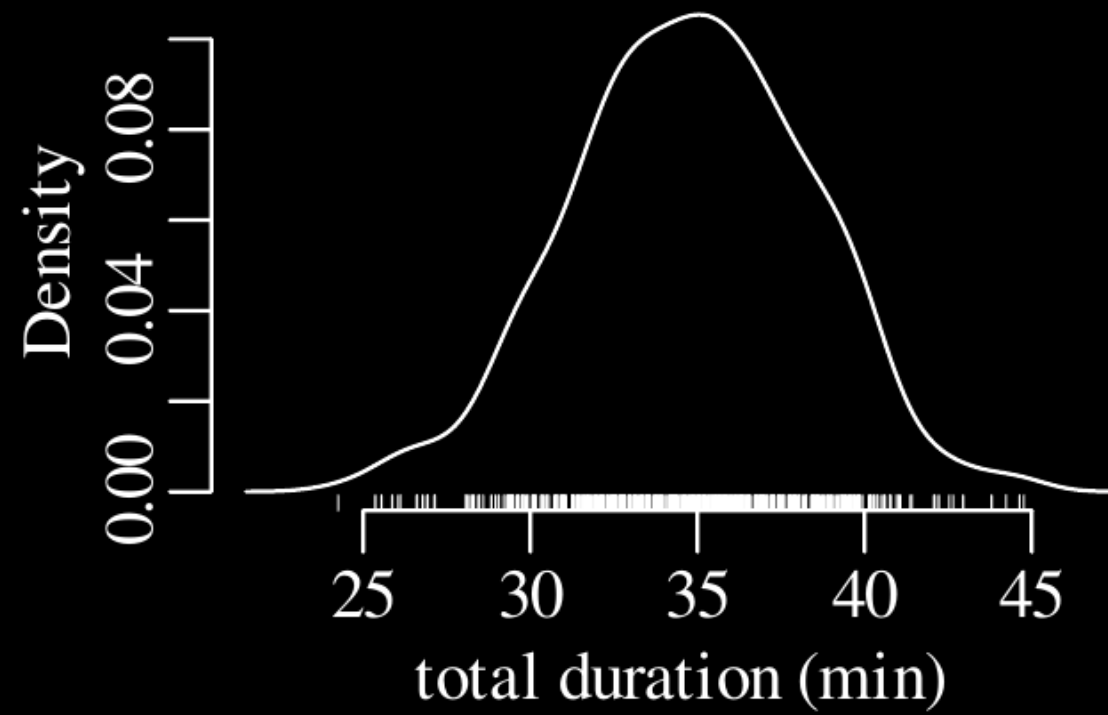
$$f(x, y | \mu_x, \mu_y, \sigma_x, \sigma_y, \sigma_{x,y}) = \frac{\exp \left(- \begin{bmatrix} (x - \mu_x) & (y - \mu_y) \end{bmatrix} \begin{bmatrix} \sigma_x^2 & \sigma_{x,y} \\ \sigma_{x,y} & \sigma_y^2 \end{bmatrix}^{-1} \begin{bmatrix} x - \mu_x \\ y - \mu_y \end{bmatrix} / 2 \right)}{2\pi \sqrt{\begin{vmatrix} \sigma_x^2 & \sigma_{x,y} \\ \sigma_{x,y} & \sigma_y^2 \end{vmatrix}}}$$

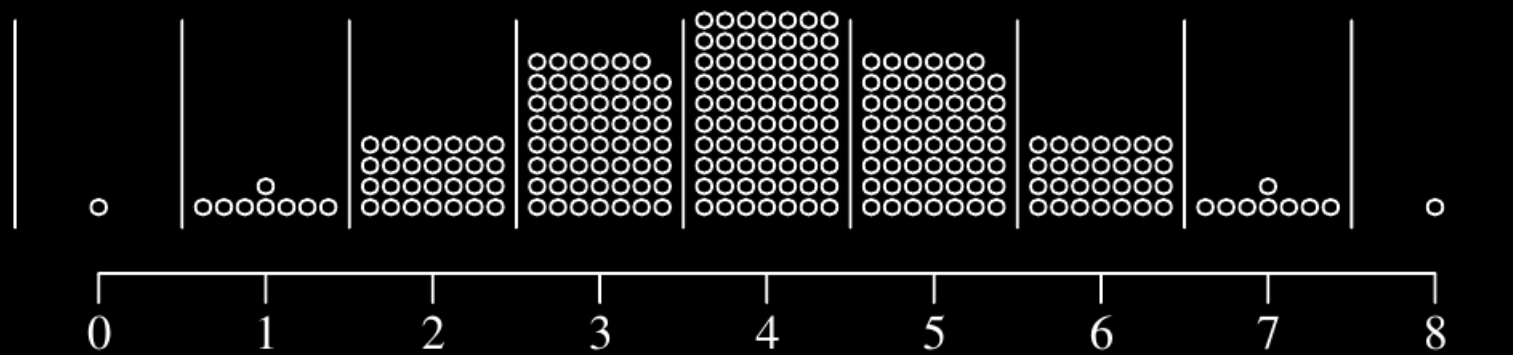
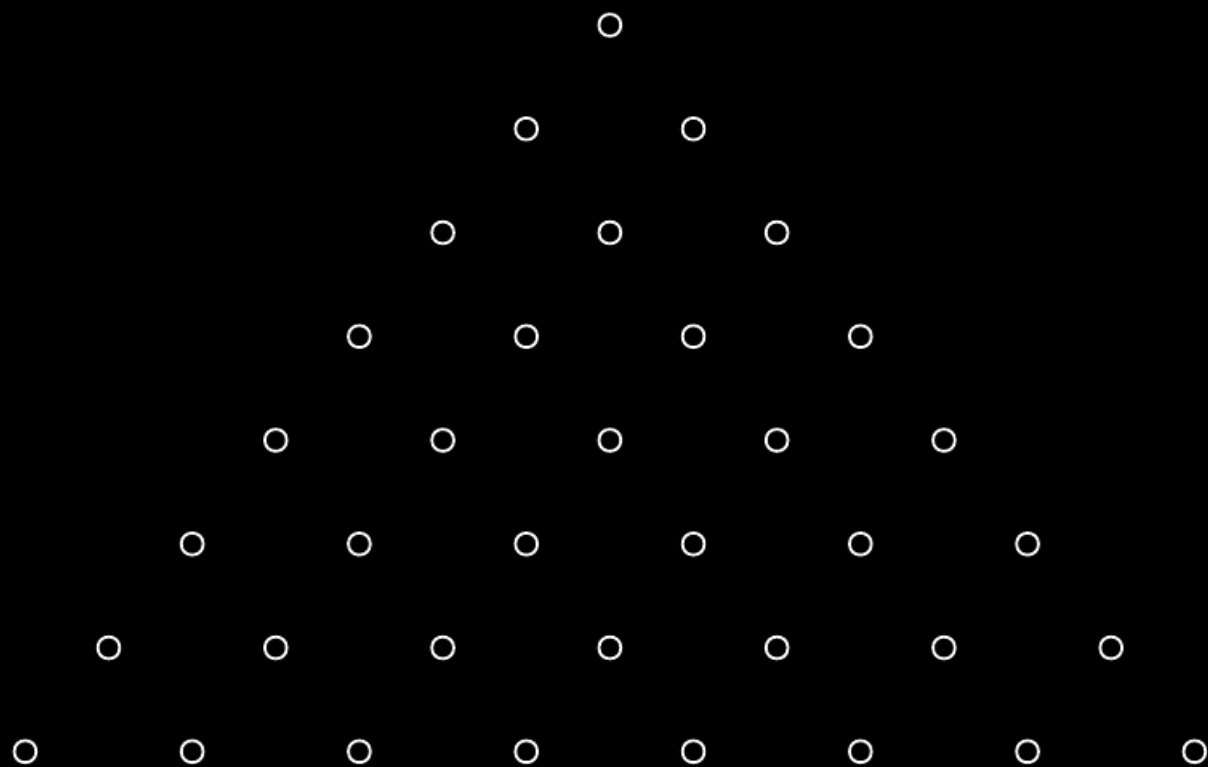


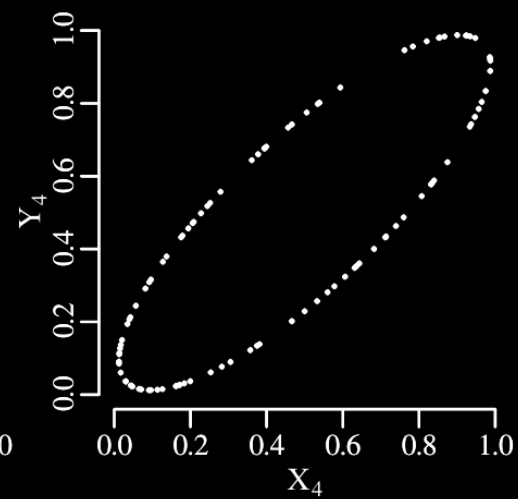
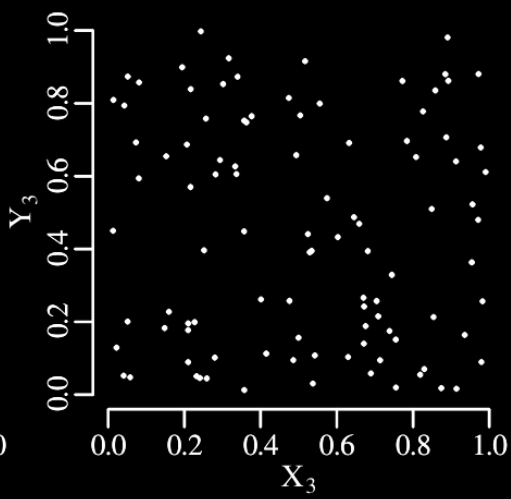
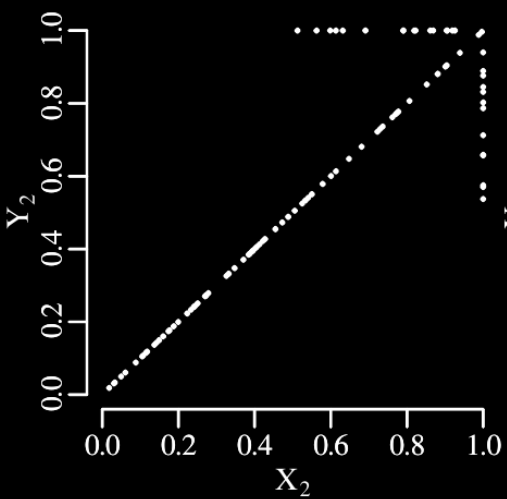
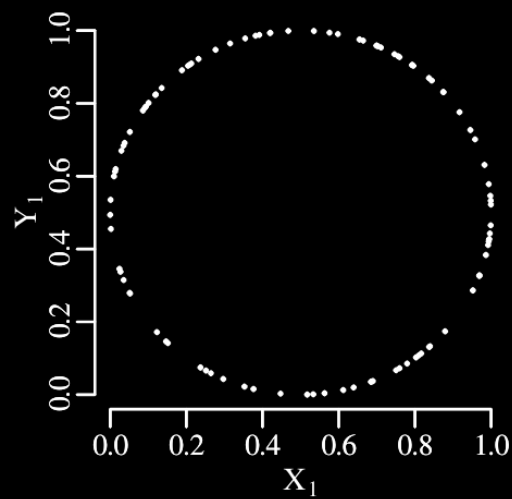
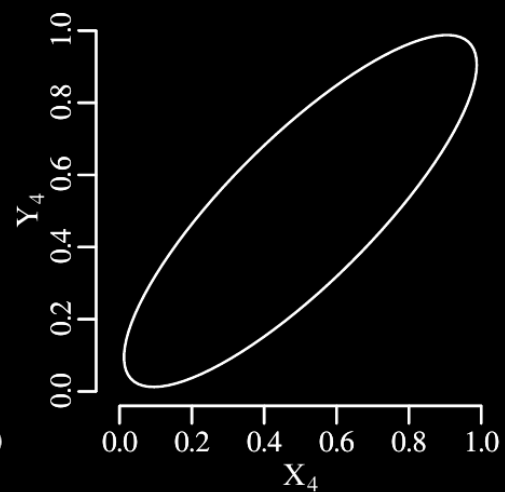
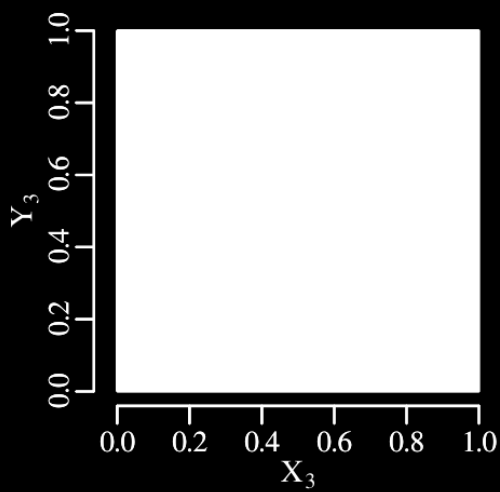
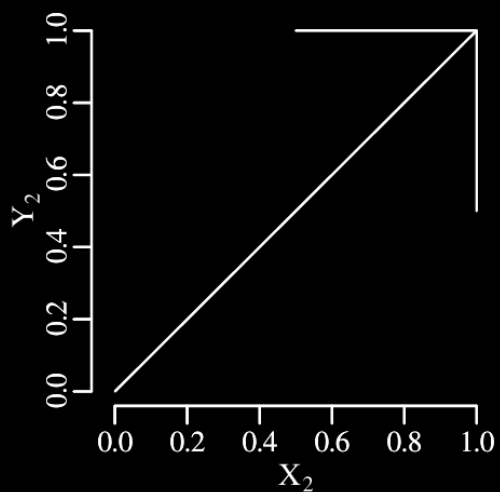
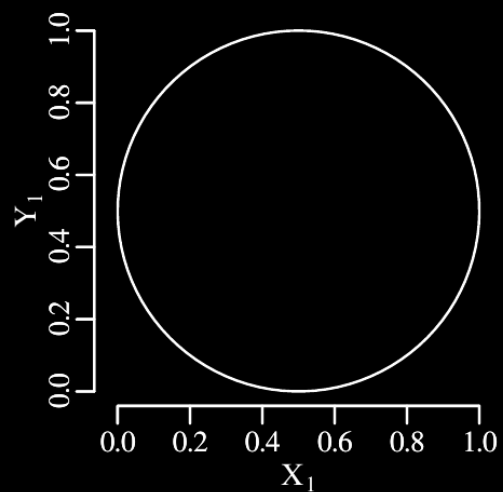


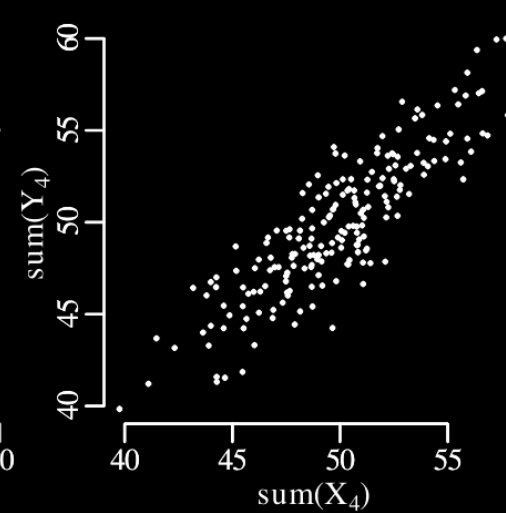
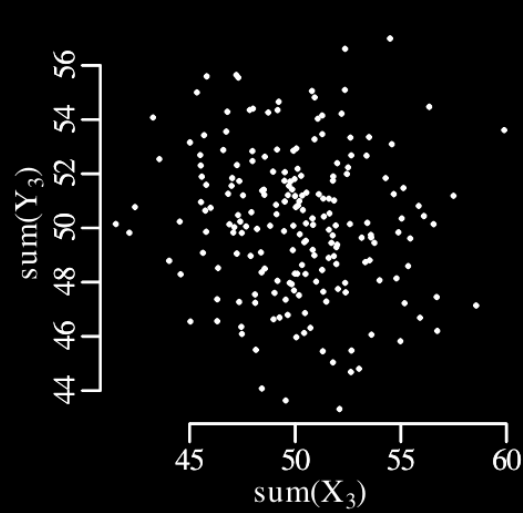
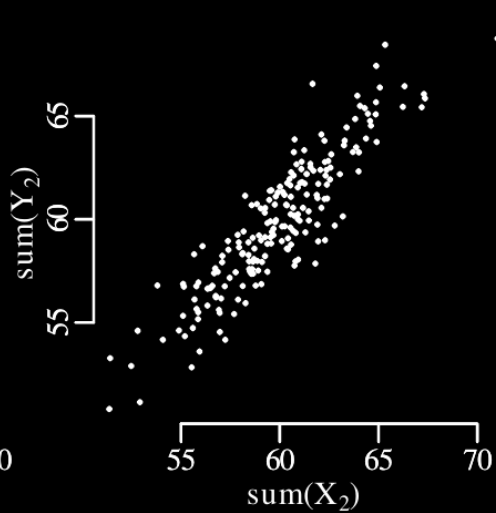
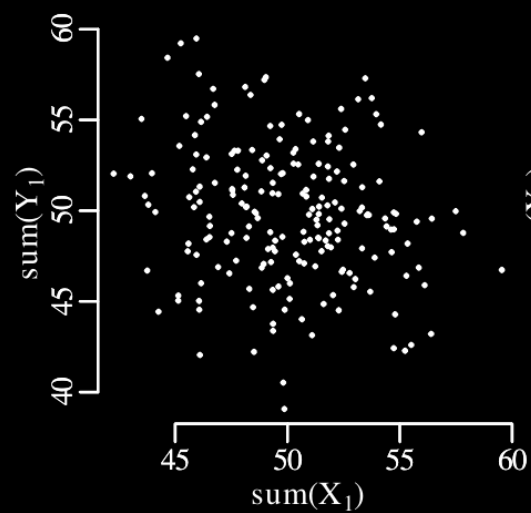
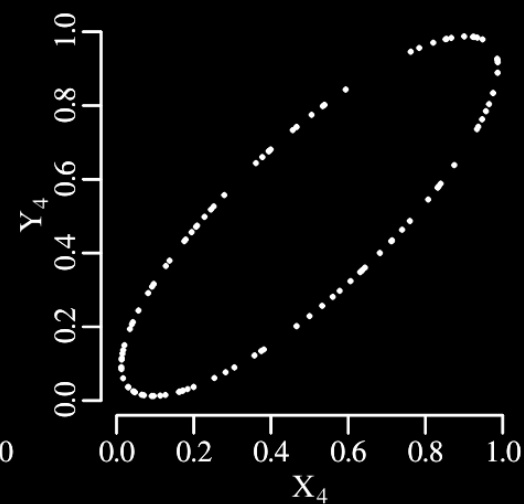
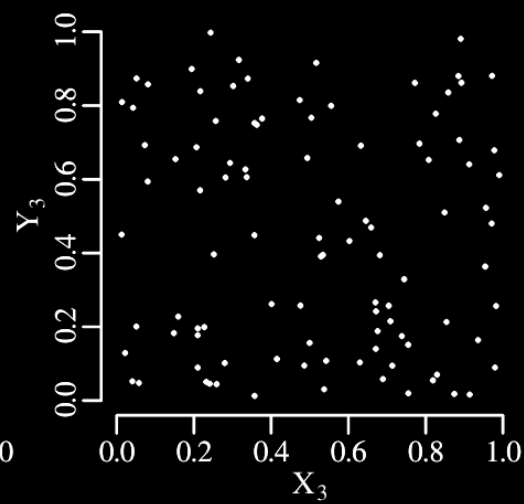
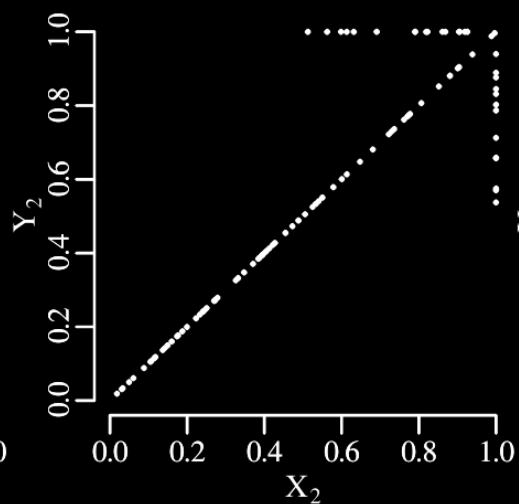
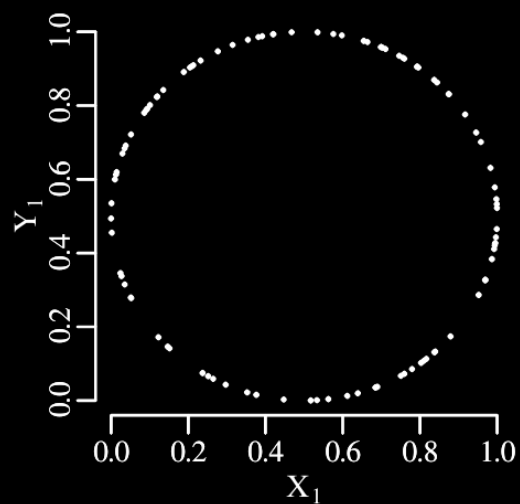












$$f(x|\mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(x - \mu)^2}{2\sigma^2}\right]$$

$$\mathcal{L}(\mu, \sigma|x_1, x_2, \dots, x_n) = \prod_{i=1}^n f(x_i|\mu, \sigma)$$

$$\begin{aligned} \mathcal{LL}(\mu, \sigma|x_1, x_2, \dots, x_n) &= \sum_{i=1}^n \ln [f(x_i|\mu, \sigma)] \\ &= \sum_{i=1}^n -\ln[\sigma] - \frac{1}{2} \ln[2\pi] - \frac{(x_i - \mu)^2}{2\sigma^2} \end{aligned}$$

$$\frac{\partial \mathcal{L}}{\partial \mu} = - \sum_{i=1}^n \frac{x_i - \mu}{\sigma^2} = 0$$

$$\Rightarrow n\mu - \sum_{i=1}^n x_i = 0$$

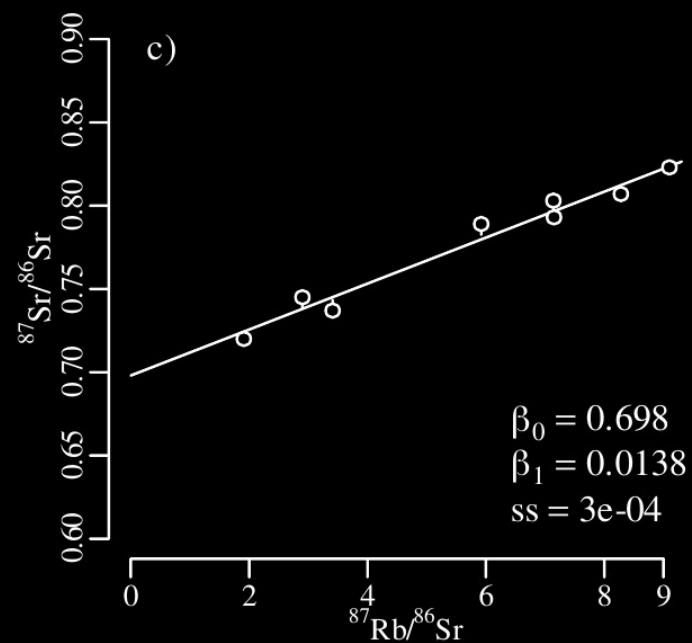
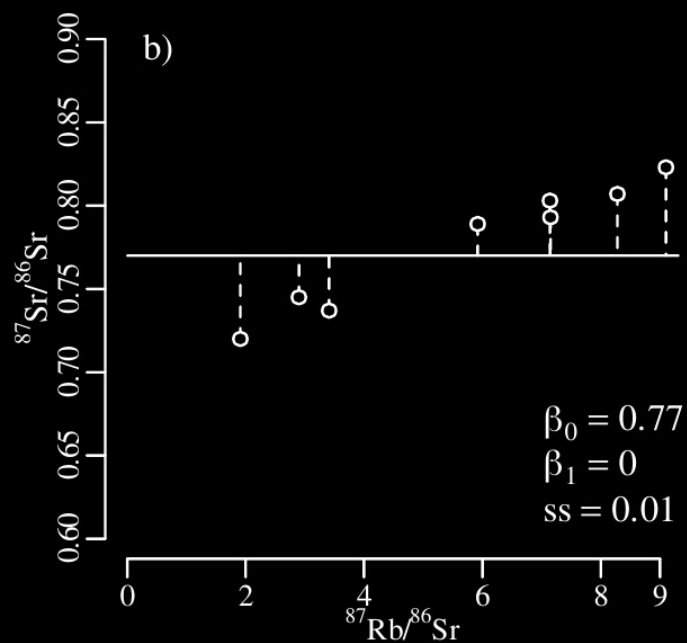
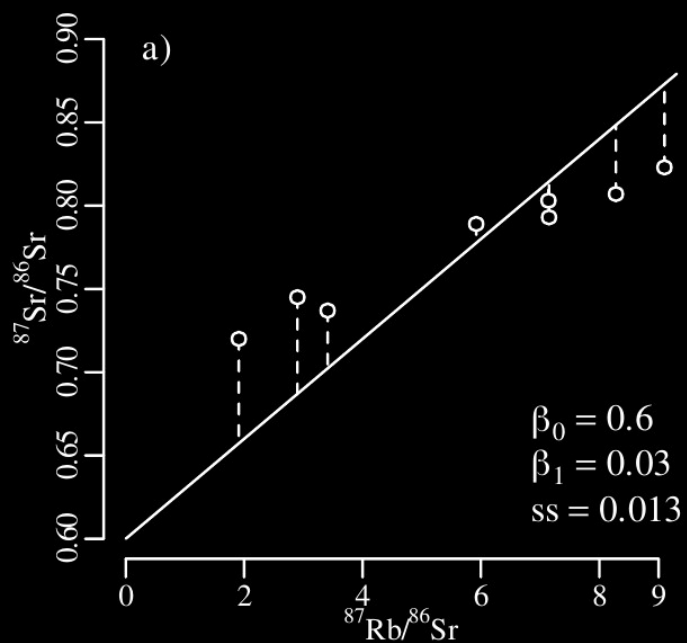
$$\Rightarrow \hat{\mu} = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\frac{\partial \mathcal{L}}{\partial \sigma} = \sum_{i=1}^n -\frac{1}{\sigma} + \frac{(x_i - \mu)^2}{\sigma^3} = 0$$

$$\Rightarrow \sum_{i=1}^n \frac{(x_i - \mu)^2}{\sigma^3} = \frac{n}{\sigma}$$

$$\Rightarrow \hat{\sigma} = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2}$$

$$ss \equiv \sum_{i=1}^n \epsilon_i^2 = \sum_{i=1}^n (\beta_0 + \beta_1 x_i - y_i)^2$$

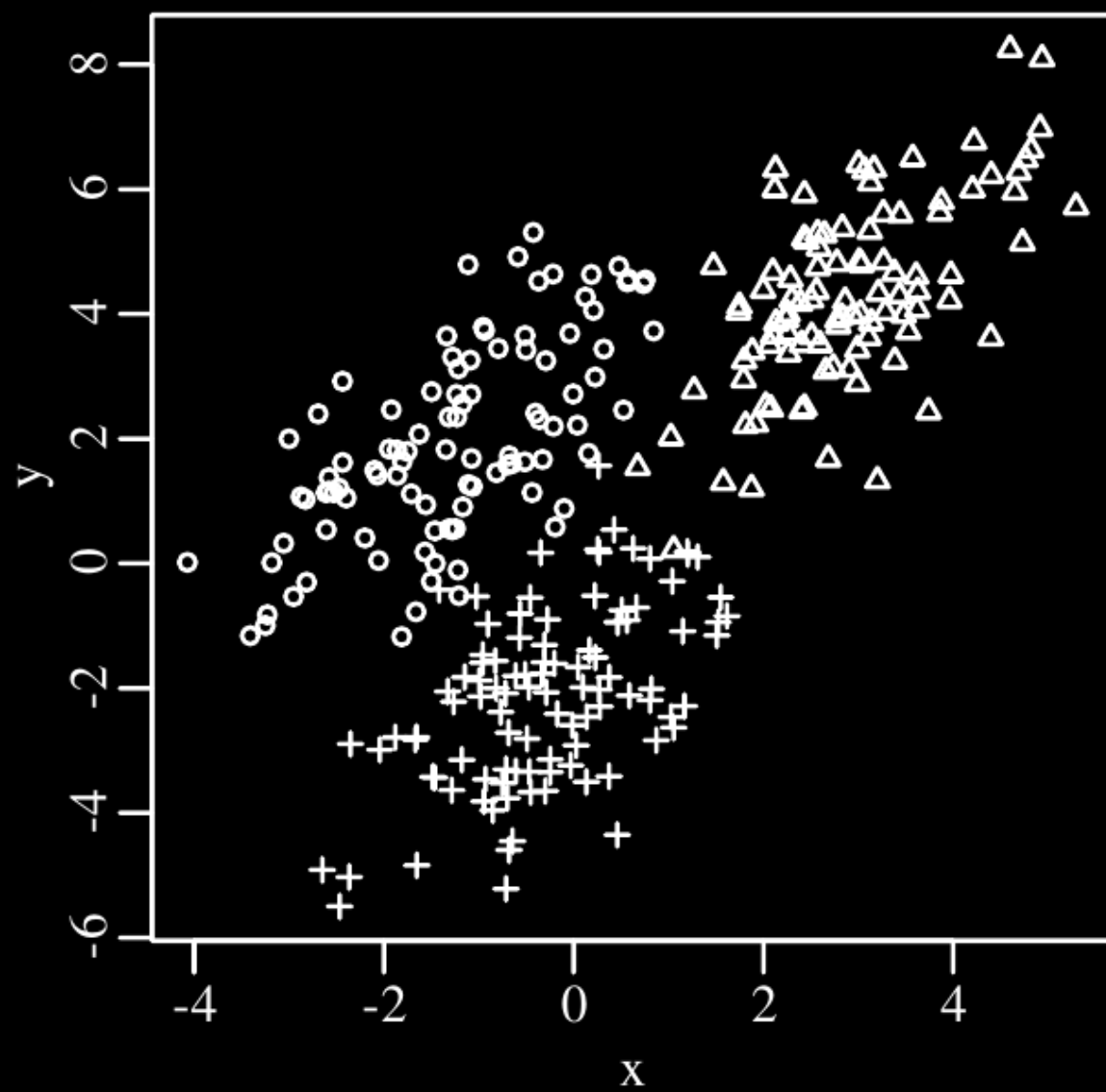


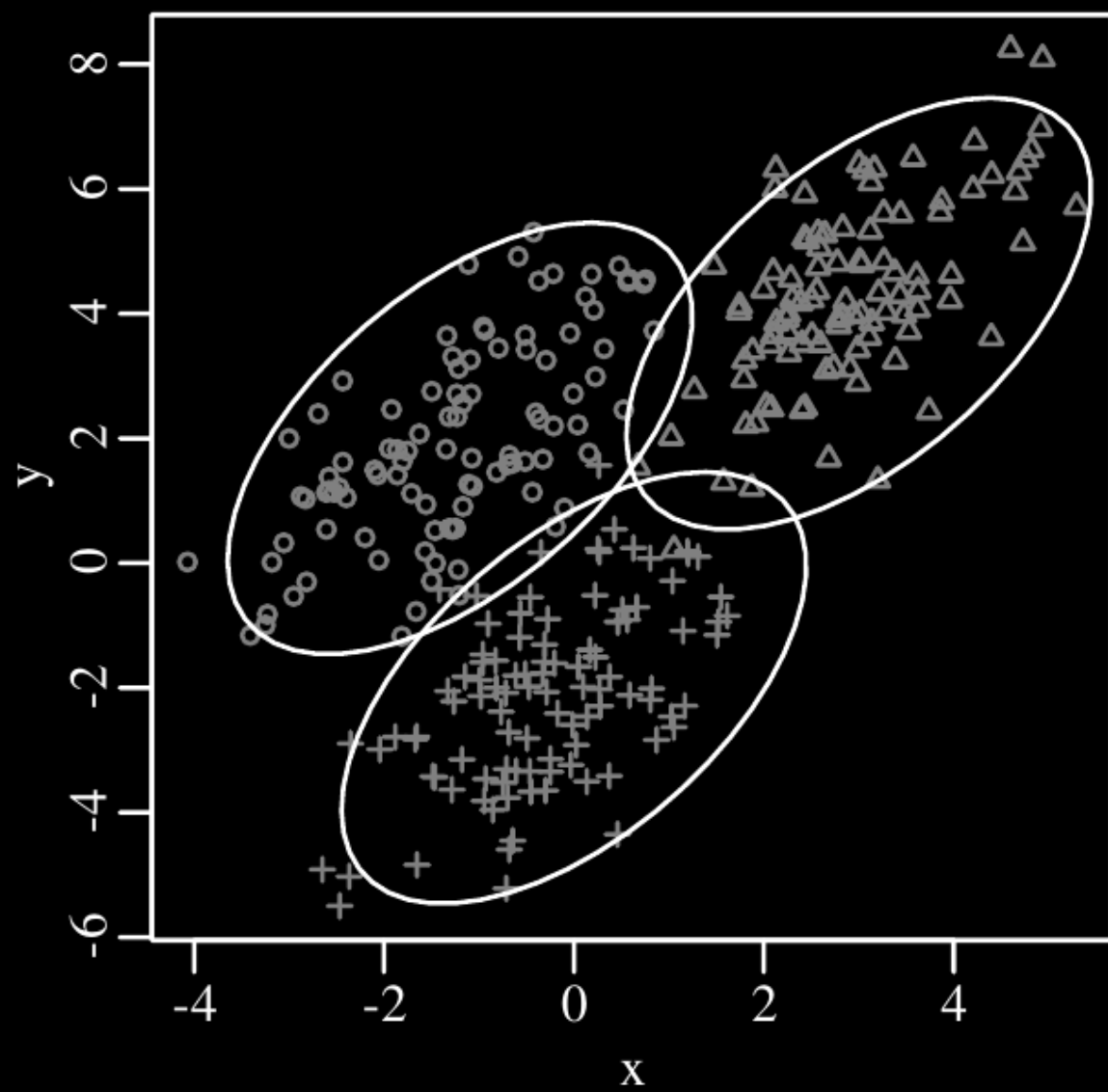
$$y_i = \beta_0 + \beta_1 x_i + \epsilon_i$$

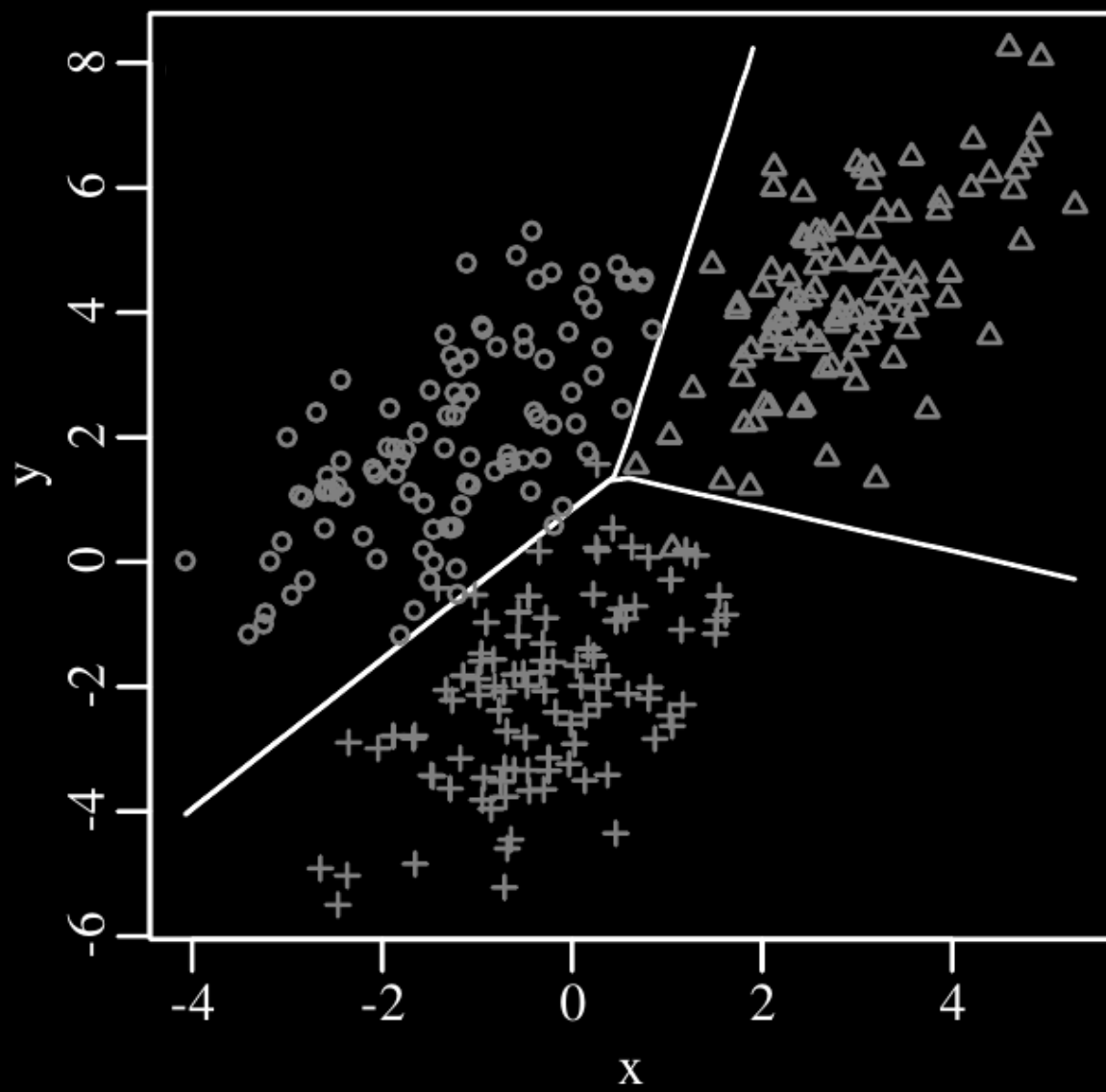
$$f(\epsilon_i|\beta_0, \beta_1, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{\epsilon_i^2}{2\sigma^2}\right], \text{ where } \epsilon_i = \beta_0 + \beta_1 x_i - y_i$$

$$\mathcal{L}(\beta_0, \beta_1, \sigma|\epsilon_1, \dots, \epsilon_n) = \prod_{i=1}^n \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{\epsilon_i^2}{2\sigma^2}\right], \text{ where } \epsilon_i = \beta_0 + \beta_1 x_i - y_i$$

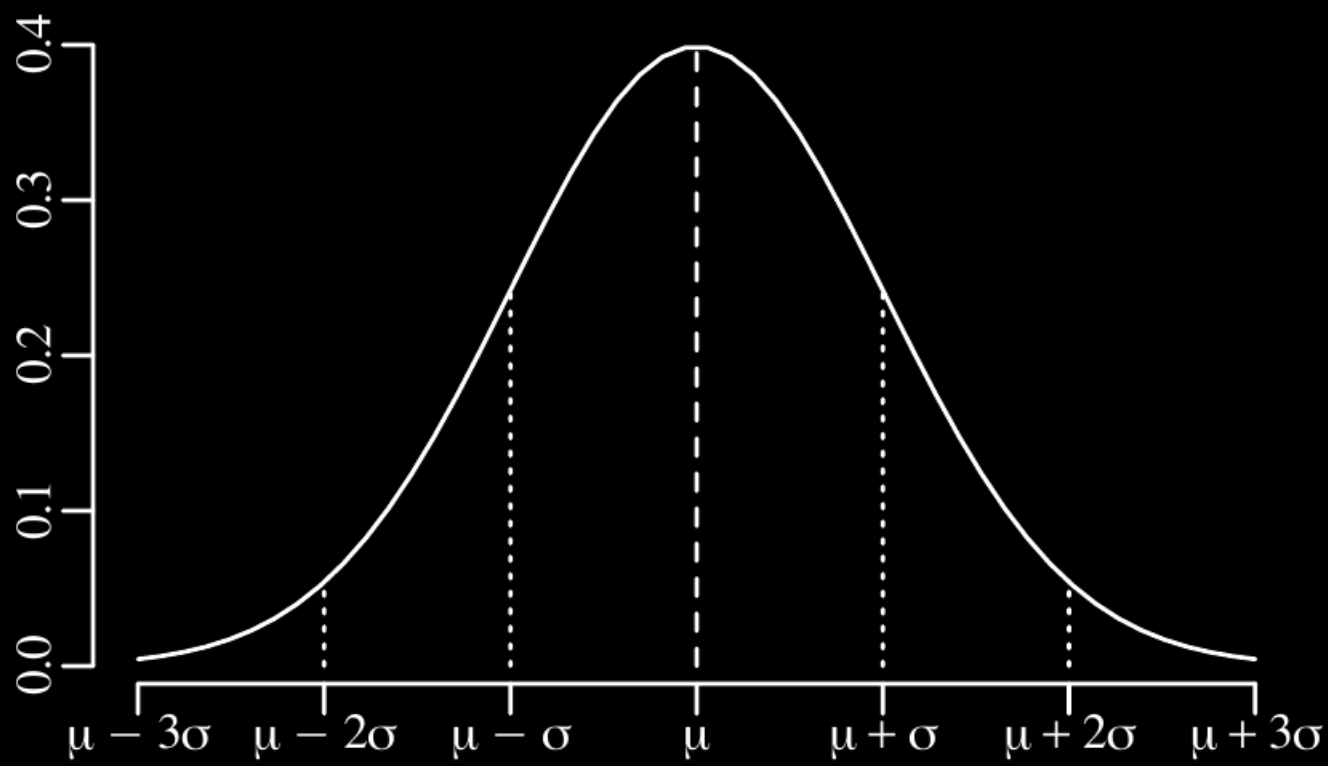
$$\begin{aligned} \max_{\beta_0, \beta_1} \left[\prod_{i=1}^n \mathcal{L} \right] &= \max_{\beta_0, \beta_1} \left[\sum_{i=1}^n \ln \mathcal{L} \right] = \max_{\beta_0, \beta_1} \left[\ln \left(\frac{1}{\sigma\sqrt{2\pi}} \right) - \sum_{i=1}^n \left(\frac{\epsilon_i^2}{2\sigma^2} \right) \right] \\ &= \max_{\beta_0, \beta_1} \left[- \sum_{i=1}^n \epsilon_i^2 \right] = \min_{\beta_0, \beta_1} \left[\sum_{i=1}^n \epsilon_i^2 \right] = \min_{\beta_0, \beta_1} (ss) \end{aligned}$$







$$f(x|\mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(x - \mu)^2}{2\sigma^2}\right]$$



	1	2	3	4	5	6	7	8	9	10
A	0.27	0.37	0.57	0.91	0.20	0.90	0.94	0.66	0.63	0.062
B	0.21	0.18	0.69	0.38	0.77	0.50	0.72	0.99	0.38	0.780

	1	2	3	4	5	6	7	8	9	10
A	0.27	0.37	0.57	0.91	0.20	0.90	0.94	0.66	0.63	0.062
B	0.21	0.18	0.69	0.38	0.77	0.50	0.72	0.99	0.38	0.780

	1	2	3	4	5	6	7	8	9	10
A/B	1.30	2.10	0.83	2.40	0.26	1.80	1.30	0.67	1.70	0.079
B/A	0.78	0.47	1.20	0.42	3.80	0.55	0.76	1.50	0.60	13.0

	1	2	3	4	5	6	7	8	9	10
A	0.27	0.37	0.57	0.91	0.20	0.90	0.94	0.66	0.63	0.062
B	0.21	0.18	0.69	0.38	0.77	0.50	0.72	0.99	0.38	0.780

	1	2	3	4	5	6	7	8	9	10
A/B	1.30	2.10	0.83	2.40	0.26	1.80	1.30	0.67	1.70	0.079
B/A	0.78	0.47	1.20	0.42	3.80	0.55	0.76	1.50	0.60	13.0
$1/(A/B)$	0.78	0.47	1.20	0.42	3.80	0.55	0.76	1.50	0.60	13.0

	1	2	3	4	5	6	7	8	9	10	
<i>A</i>	0.27	0.37	0.57	0.91	0.20	0.90	0.94	0.66	0.63	0.062	
<i>B</i>	0.21	0.18	0.69	0.38	0.77	0.50	0.72	0.99	0.38	0.780	
	1	2	3	4	5	6	7	8	9	10	mean
<i>A/B</i>	1.30	2.10	0.83	2.40	0.26	1.80	1.30	0.67	1.70	0.079	1.20
<i>B/A</i>	0.78	0.47	1.20	0.42	3.80	0.55	0.76	1.50	0.60	13.0	2.30
<i>1/(A/B)</i>	0.78	0.47	1.20	0.42	3.80	0.55	0.76	1.50	0.60	13.0	

$$\frac{1}{\overline{A/B}} = \frac{1}{1.20} = 0.81 \neq 2.30 = \overline{B/A}$$

$$\text{and } \frac{1}{\overline{B/A}} = \frac{1}{2.30} = 0.44 \neq 1.20 = \overline{A/B}$$

	1	2	3	4	5	6	7	8	9	10
A	0.27	0.37	0.57	0.91	0.20	0.90	0.94	0.66	0.63	0.062
B	0.21	0.18	0.69	0.38	0.77	0.50	0.72	0.99	0.38	0.780

	1	2	3	4	5	6	7	8	9	10	mean
$\ln[A/B]$	0.25	0.75	-0.18	0.86	-1.30	0.59	0.27	-0.41	0.50	-2.50	-0.12
$\ln[B/A]$	-0.25	-0.75	0.18	-0.86	1.30	-0.59	-0.27	0.41	-0.50	2.50	0.12

	mean	exp[mean]
$\ln[A/B]$	-0.12	0.88
$\ln[B/A]$	0.12	1.13

	1	2	3	4	5	6	7	8	9	10
A	0.27	0.37	0.57	0.91	0.20	0.90	0.94	0.66	0.63	0.062
B	0.21	0.18	0.69	0.38	0.77	0.50	0.72	0.99	0.38	0.780

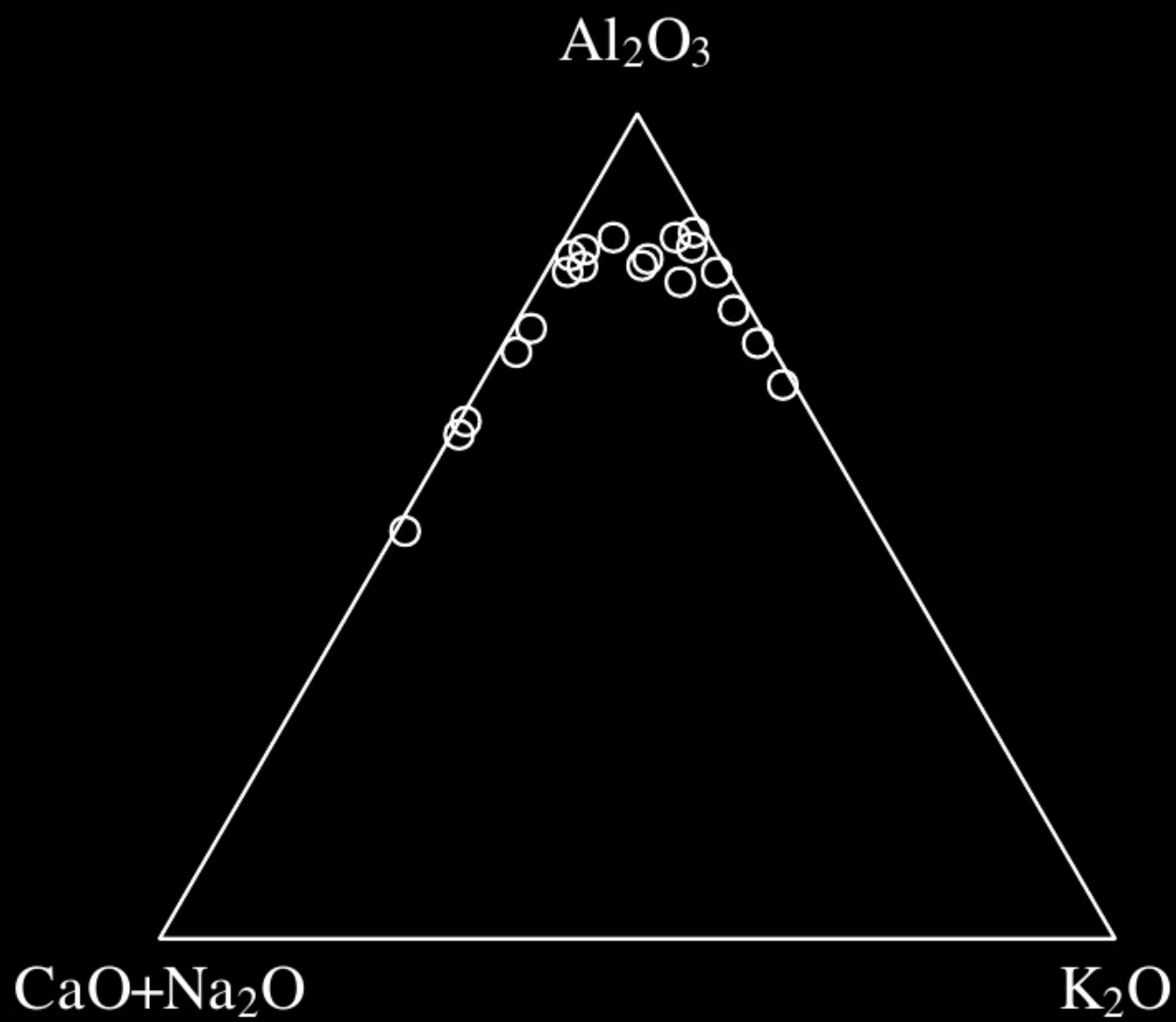
	1	2	3	4	5	6	7	8	9	10	mean
$\ln[A/B]$	0.25	0.75	-0.18	0.86	-1.30	0.59	0.27	-0.41	0.50	-2.50	-0.12
$\ln[B/A]$	-0.25	-0.75	0.18	-0.86	1.30	-0.59	-0.27	0.41	-0.50	2.50	0.12

	mean	exp[mean]
$\ln[A/B]$	-0.12	0.88
$\ln[B/A]$	0.12	1.13

$$\frac{1}{g(A/B)} = \frac{1}{0.88} = 1.13 = g(B/A)$$

and

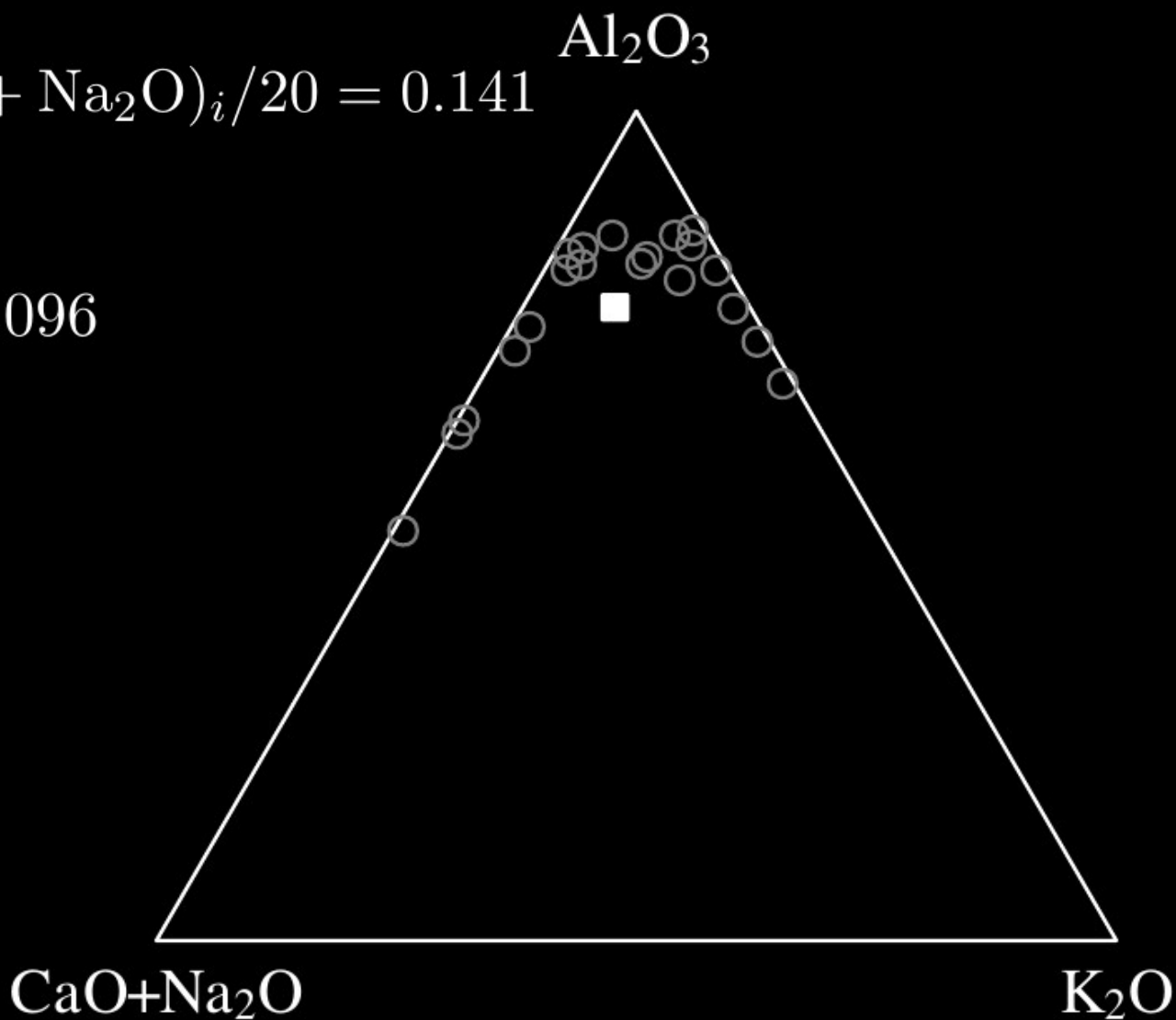
$$\frac{1}{g(B/A)} = \frac{1}{1.13} = 0.88 = g(A/B)$$



$$\overline{\text{Al}_2\text{O}_3} = \sum_{i=1}^{20} (\text{Al}_2\text{O}_3)_i / 20 = 0.763$$

$$\overline{\text{CaO} + \text{Na}_2\text{O}} = \sum_{i=1}^{20} (\text{CaO}_2 + \text{Na}_2\text{O})_i / 20 = 0.141$$

$$\overline{\text{K}_2\text{O}} = \sum_{i=1}^{20} (\text{K}_2\text{O})_i / 20 = 0.096$$



$$s[\text{Al}_2\text{O}_3] = \sqrt{\sum_{i=1}^{20} \frac{((\text{Al}_2\text{O}_3)_i - 0.763)^2}{19}} = 0.0975$$

$$s[\text{CaO} + \text{Na}_2\text{O}] = \sqrt{\sum_{i=1}^{20} \frac{((\text{CaO}_2 + \text{Na}_2\text{O})_i - 0.141)^2}{19}} = 0.142$$

$$s[\text{K}_2\text{O}] = \sqrt{\sum_{i=1}^{20} \frac{((\text{K}_2\text{O})_i - 0.096)^2}{19}} = 0.0926$$

$$\text{Al}_2\text{O}_3 : 0.763 \pm 0.195$$

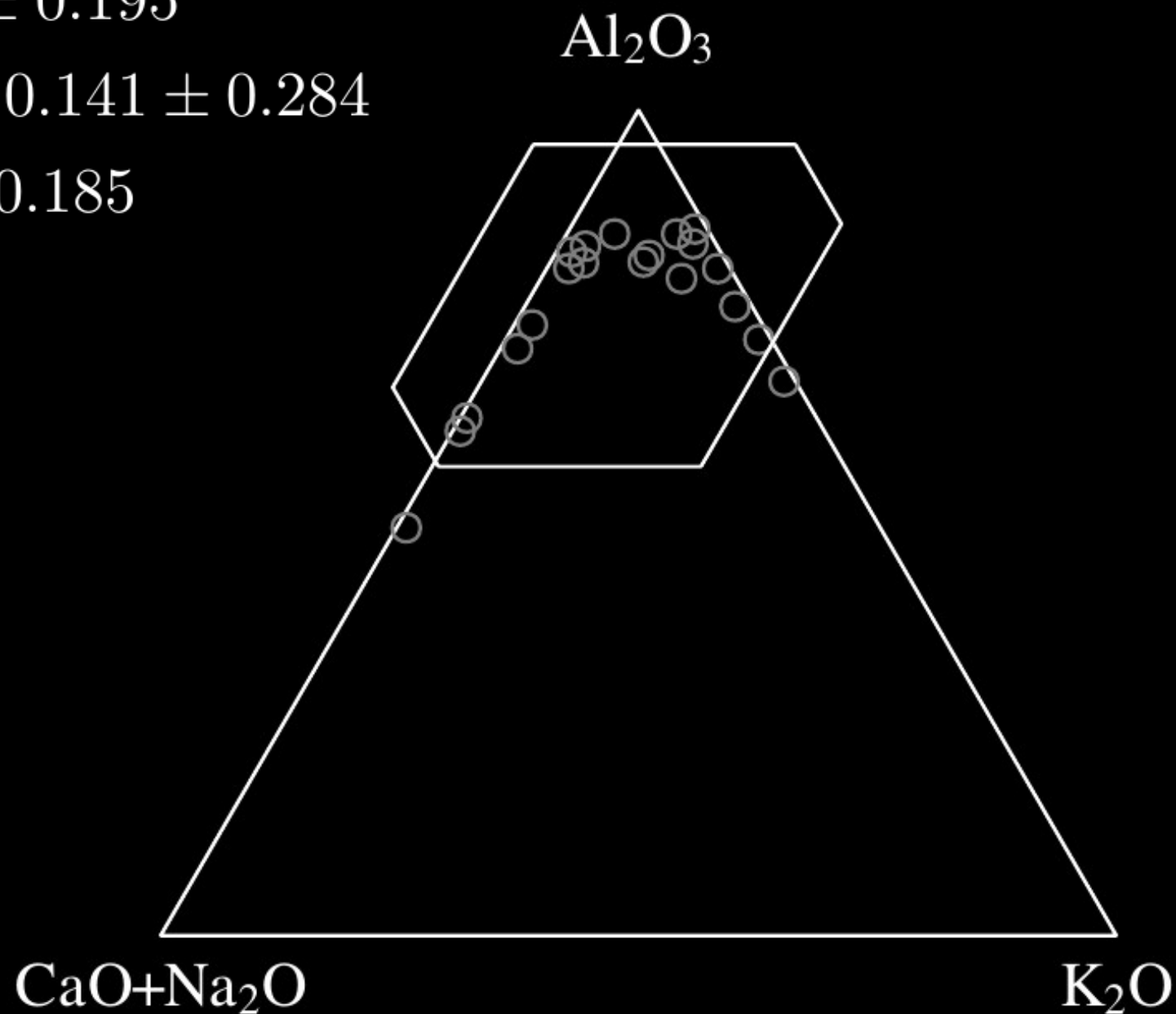
$$\text{CaO} + \text{Na}_2\text{O} : 0.141 \pm 0.284$$

$$\text{K}_2\text{O} : 0.096 \pm 0.185$$

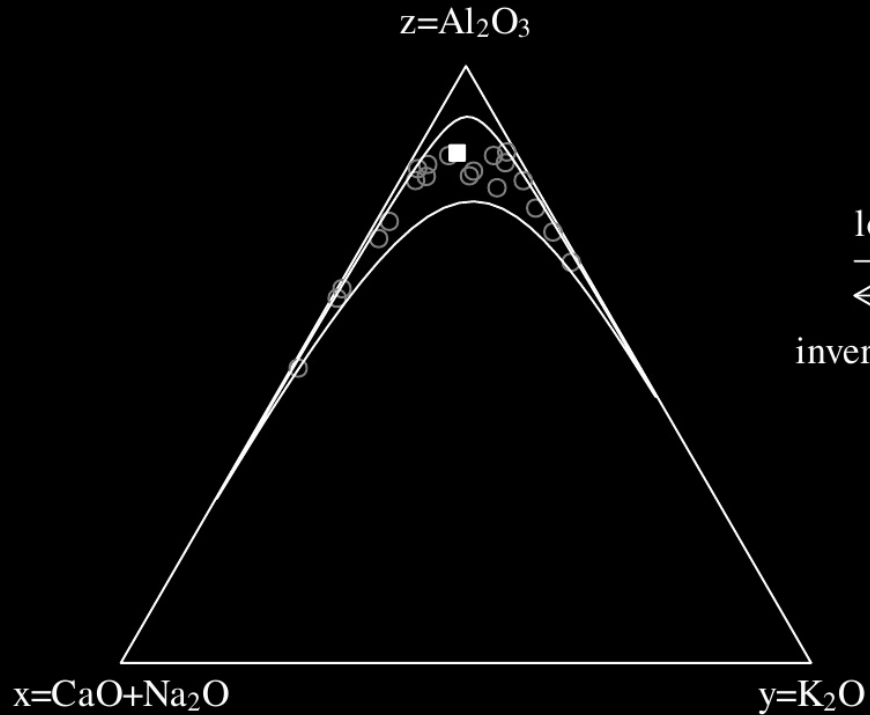
$\text{Al}_2\text{O}_3 : 0.763 \pm 0.195$

$\text{CaO} + \text{Na}_2\text{O} : 0.141 \pm 0.284$

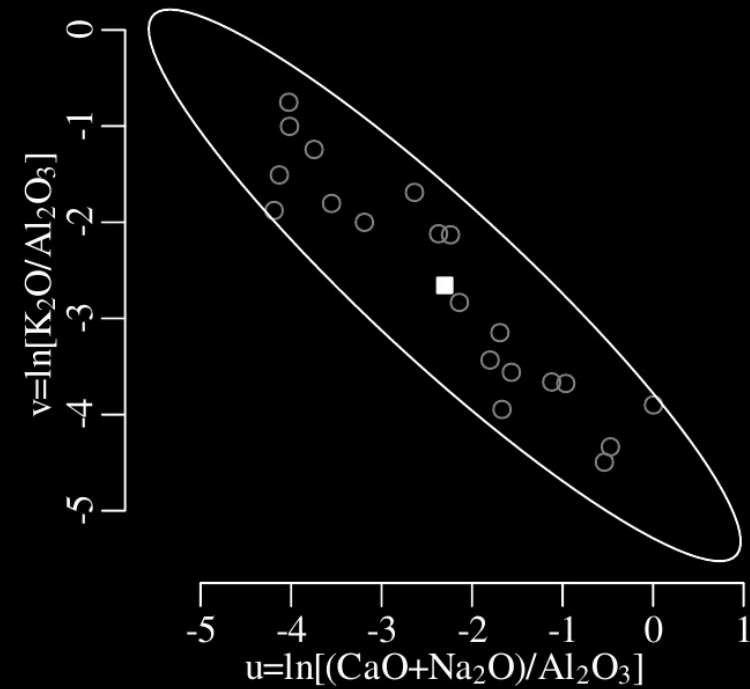
$\text{K}_2\text{O} : 0.096 \pm 0.185$



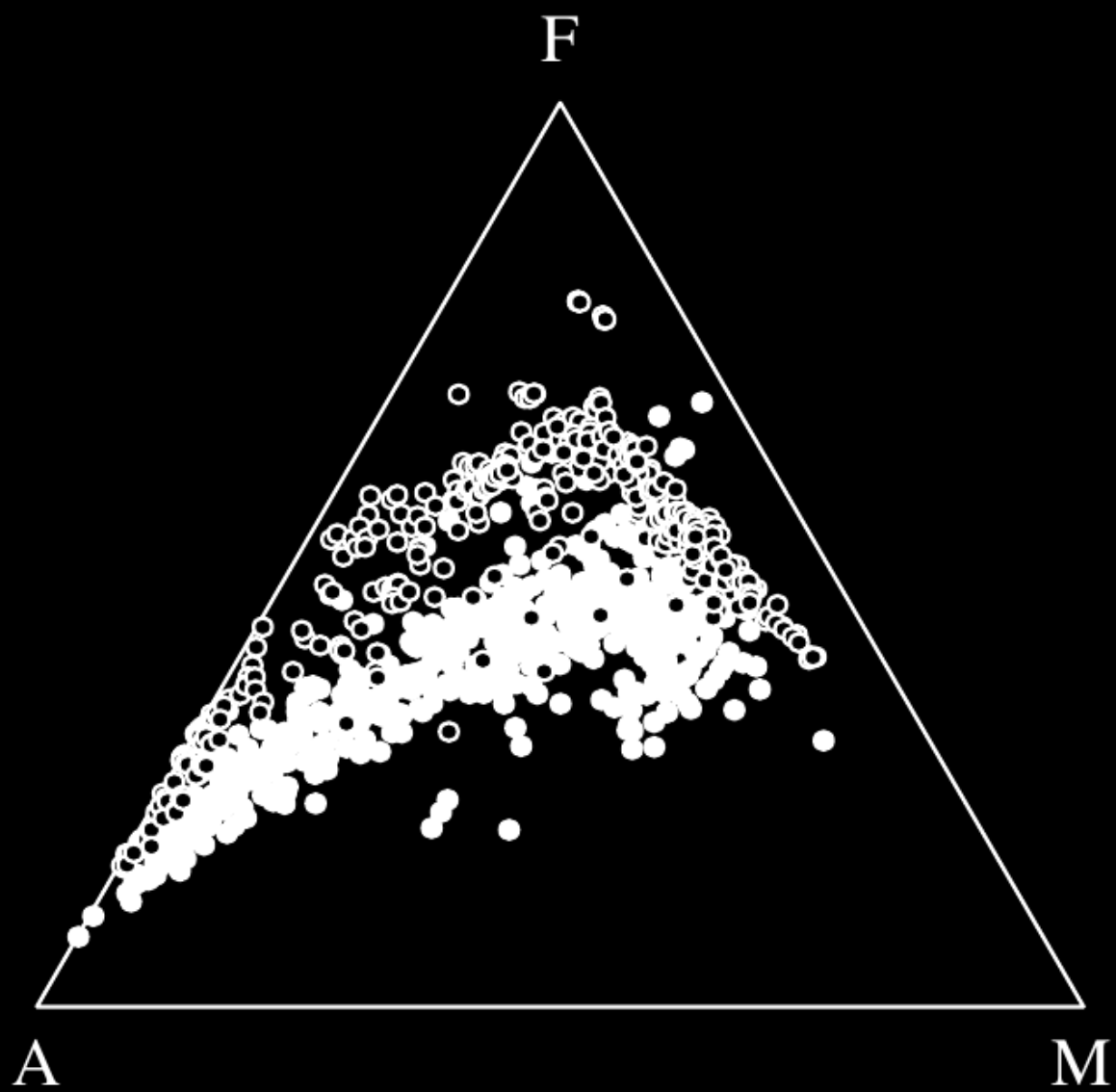
$$v = \ln \left[\frac{x}{z} \right], \quad w = \ln \left[\frac{y}{z} \right]$$

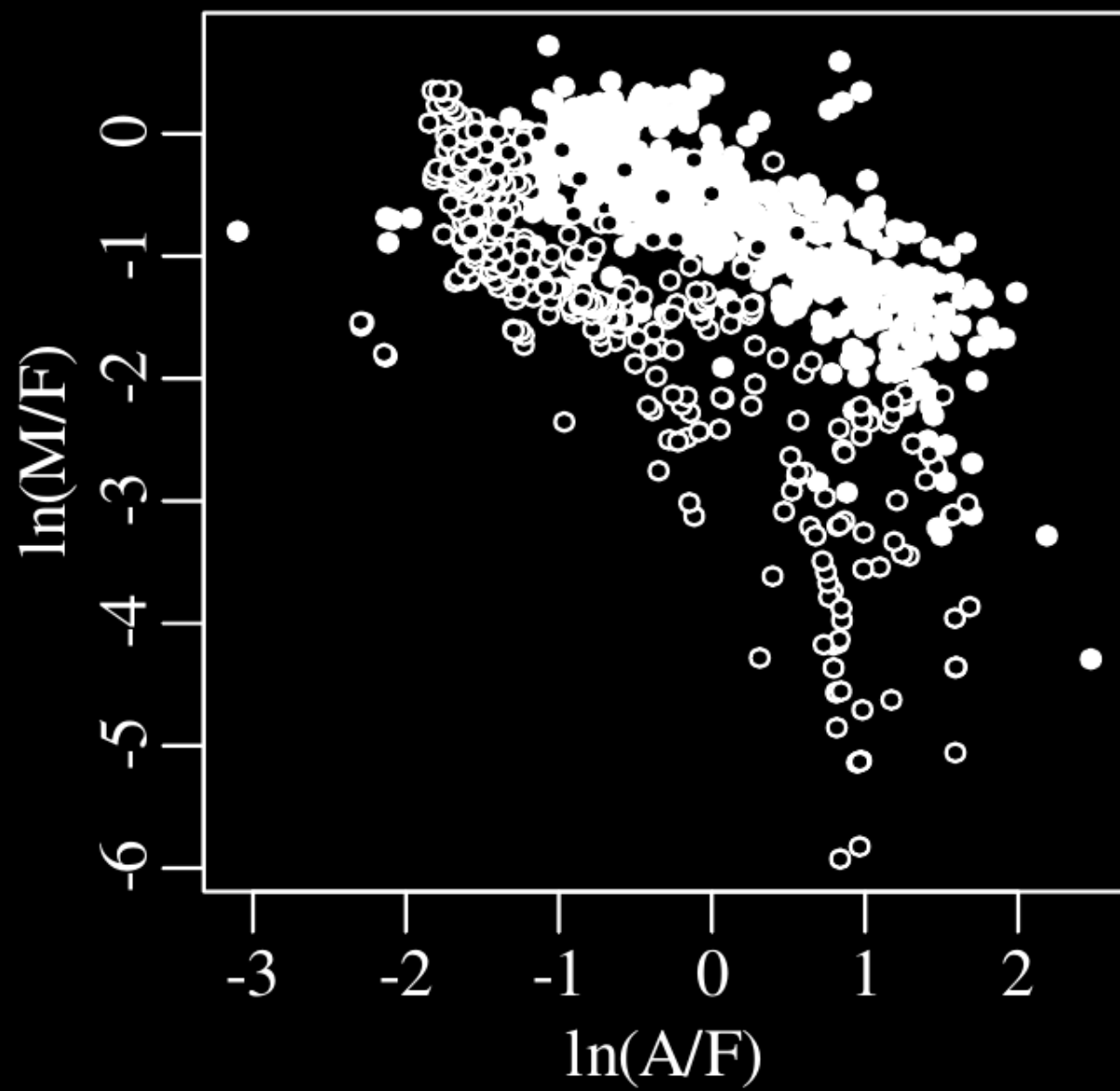


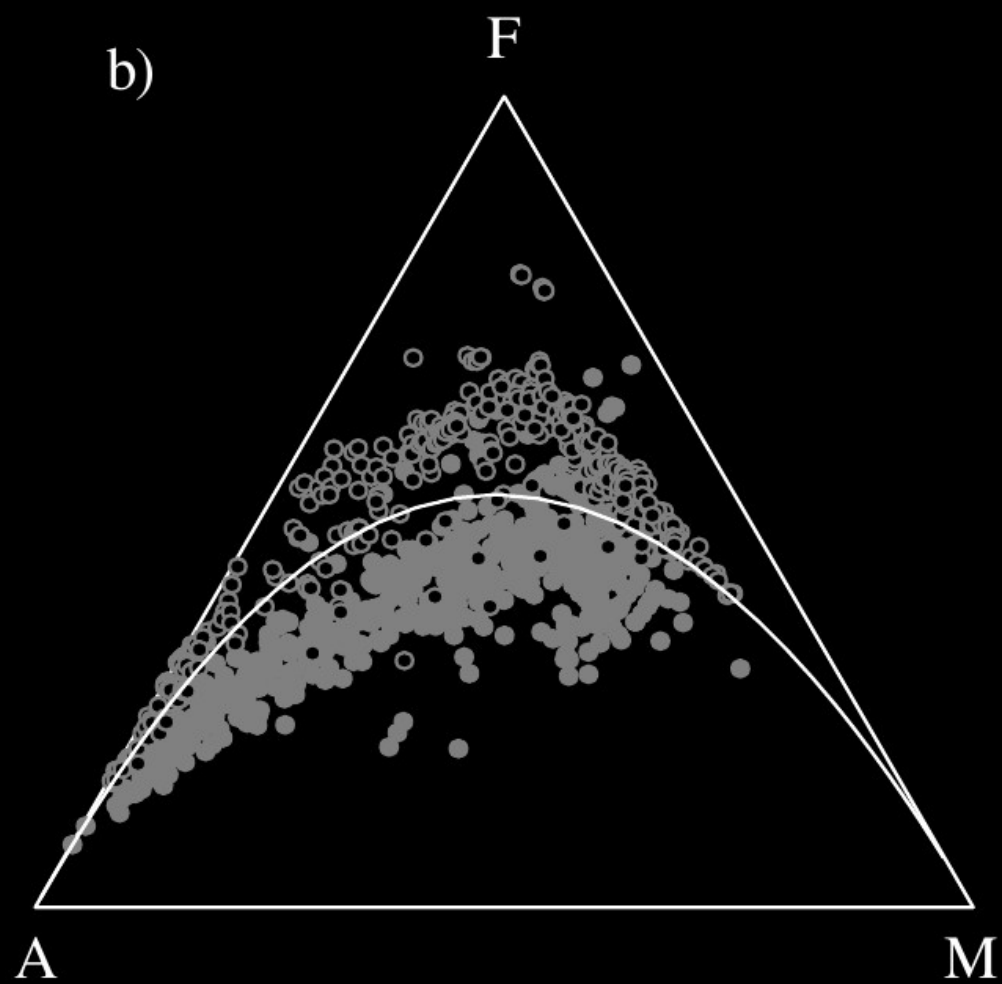
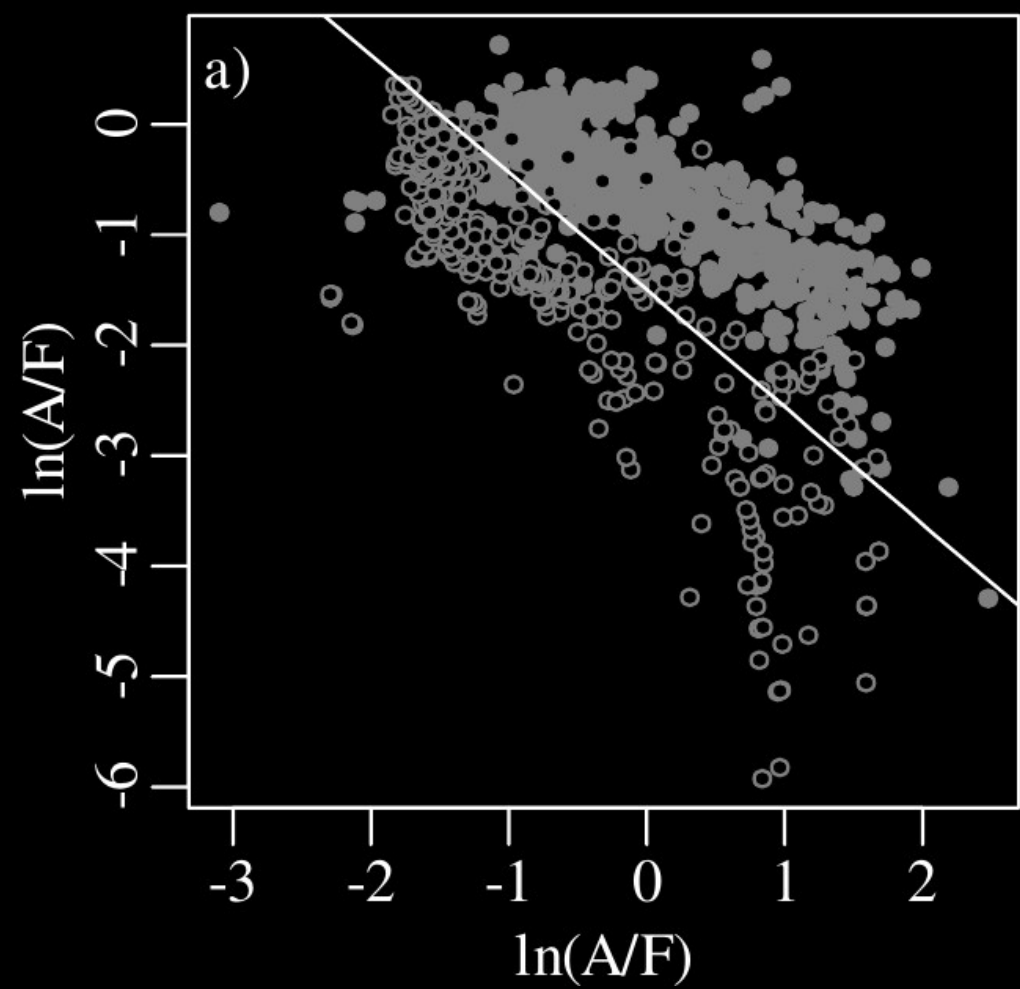
$\xrightarrow{\text{logratio transformation}}$
 $\xleftarrow{\text{inverse logratio transformation}}$



$$x = \frac{\exp[v]}{\exp[v] + \exp[w] + 1}, \quad y = \frac{\exp[w]}{\exp[v] + \exp[w] + 1}, \quad z = \frac{1}{\exp[v] + \exp[w] + 1}$$







1. `mtcars` is one of R's built-in datasets. It contains a table with the fuel consumption and 10 aspects of automobile design and performance for 32 automobiles features in a 1974 edition of 'Motor Trend' magazine.

- (a) Calculate the average fuel consumption of these 32 vehicles –which are listed in miles per gallon– using the arithmetic mean.
- (b) Convert the data to European units, namely litres per 100km. To convert from mpg to litre/100km, use the following formula:

$$1 \text{ litre}/100\text{km} = \frac{235.21}{1 \text{ mpg}}$$

- (c) Calculate the arithmetic mean fuel consumption in litre/100km.
- (d) Convert the arithmetic mean number of miles per gallon (from step 1a) to units of litre/100km. How does the resulting value compare with that obtained from step 1c.
- (e) Compute the geometric mean fuel consumption in mpg and litre/100km. Then convert the units of these mean values and compare with the result observed under step 1d.

2. Load `geostats`' test data into memory and plot the CaO–K₂O–Na₂O **subcomposition** on a ternary diagram. Add the arithmetic and geometric mean compositions to the plot. Experiment with the arguments to the `ternary` function to make the plot look as clear as possible.