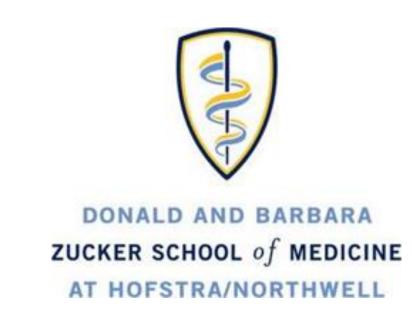
# Using org-mode for scientific posters

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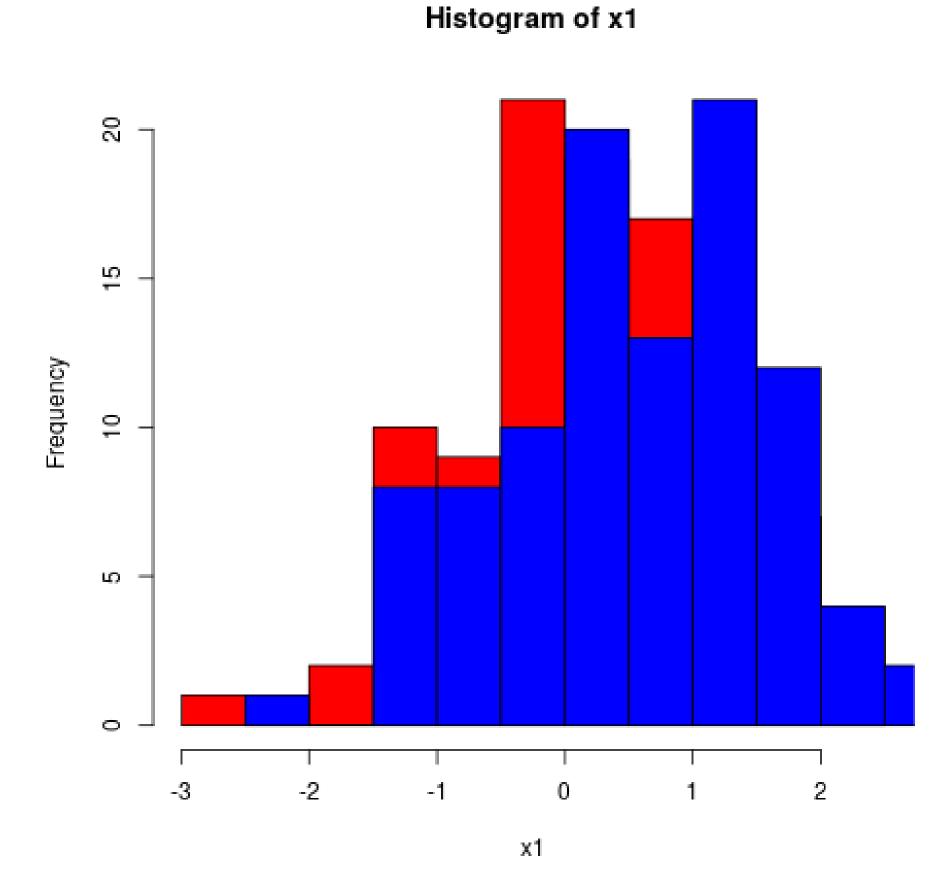
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#### Background

- Here we show how org-mode (version 9.1.7) and emacs (version 25.2.2) can be used to make decent looking scientific posters
- · With org-mode we can populate the poster with code, graphs and numbers from inline code in languages such as R, python, Matlab and even shell scripting
- For example, this poster was created on 2018-05-01 00:19 on Ubuntu 17.10.
- Inline code could look like this (which will produce a graph; Fig. 1):

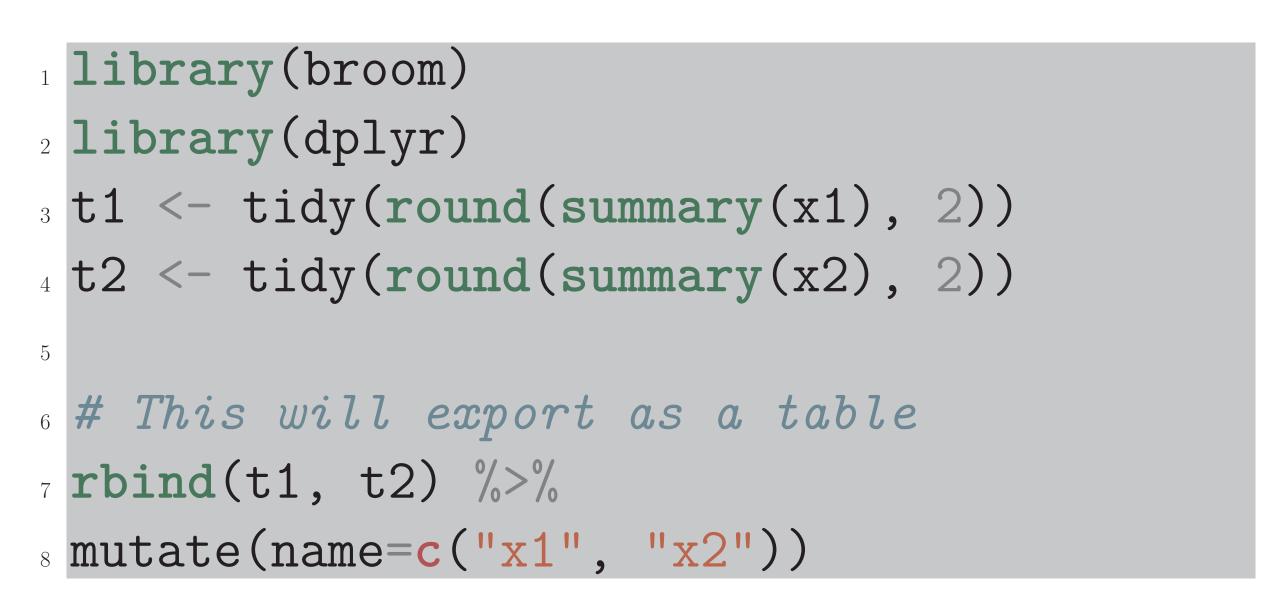
```
set.seed(20180402)
2 \times 1 <- rnorm(100, 0, 1)
_3 x2 <- rnorm(100, 0.5, 1)
hist(x1, col="red")
5 hist(x2, col="blue", add=TRUE)
```



**Figure 1:** This is the output.

## Inline code and tables

- Tables are powerful in org-mode and even include spreadsheet capabilities
- Some code to process the first vector from above to make a table out of its summary could look like this, which would result in a little table (Table 1):



q1 median mean q3 maximum name minimum -2.29 -0.49 0.11 0.14 0.8 2.47 X1 0.07 0.13 0.85 -2.17 -0.45 2.23 X2 **Table 1:** A table summarizing the two distributions.

### Graphics

- · We can use shell scripting to grab an image with curl from the internet (Fig. 2):
  - # Download emacs icon from gnu.org
  - curl -0 https://www.gnu.org/software/emacs/images/emacs.png



Figure 2: This is the downloaded image.

#### Math

 Let's describe how to compute the distance between the two simulated distributions x1 and x2 from before:

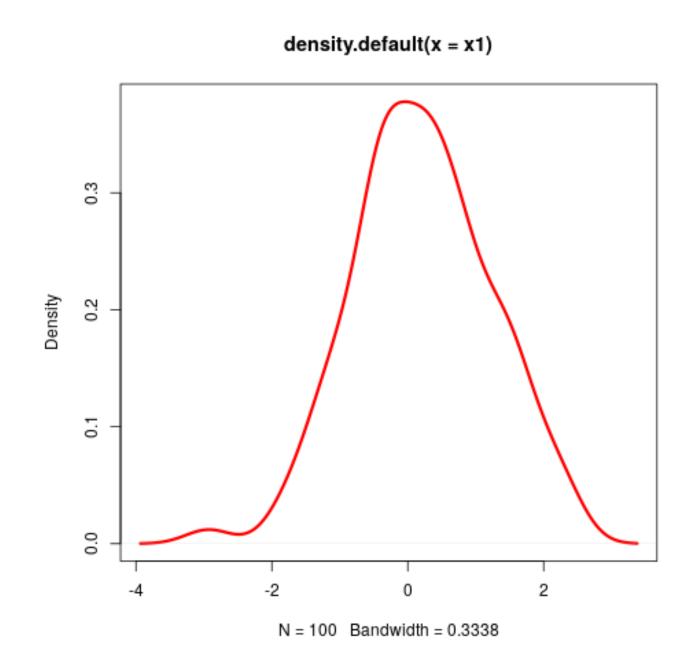
The Kullback-Leibler (KL) divergence measures the difference between two probability distributions (i.e., the loss of information when one distribution is used to approximate another). The KL divergence is thus defined as

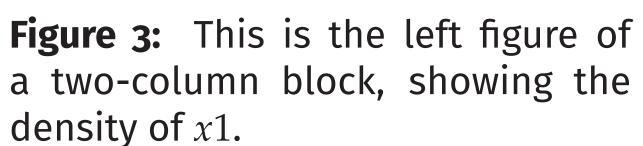
$$D_{KL}(P||Q) = \sum_{i=1}^{n} P(i) \log \frac{P(i)}{Q(i)}$$
 (1)

with P and Q being two probability distribution functions and n the number of sample points. Since  $D_{KL}(P||Q)$  is not equal to  $D_{KL}(Q||P)$ , a symmetric variation of the KL divergence can be derived as follows:

$$D_{KL}(P,Q) = \sum_{i=1}^{n} \left( P(i) \log \frac{P(i)}{Q(i)} + Q(i) \log \frac{Q(i)}{P(i)} \right).$$
 (2)

#### Columns





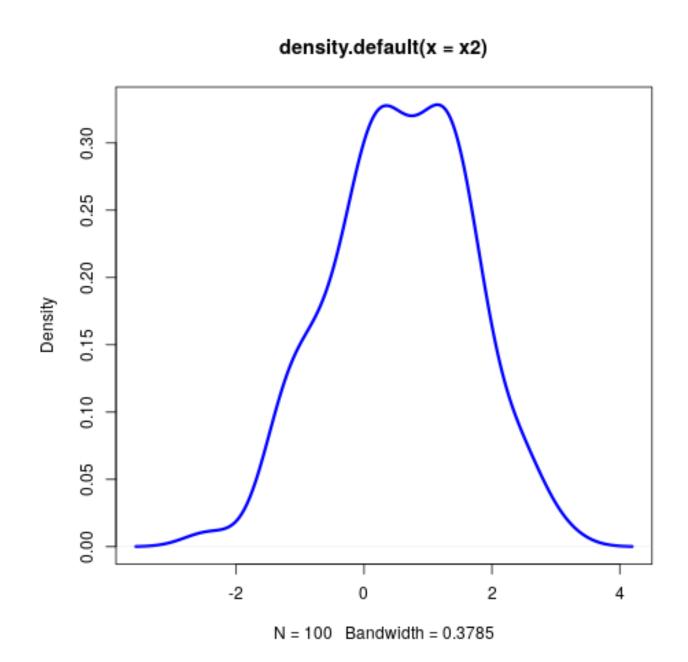


Figure 4: This is the right figure. It shows the density of x2.

- This example is meant to show how versatile org-mode is
- Scientific posters can be produced with a simple text editor