## EM algorithm A: The EM algorithm for the t-distribution

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## The marginal distribution of X is the t-distribution

The joint density of Y = (X, W) is given by

$$f(x,w) = \frac{1}{\sqrt{\pi\nu\sigma^2}2^{(\nu+1)/2}\Gamma(\nu/2)}w^{(\nu-1)/2}\exp\left(-\frac{w}{2}\left(1 + \frac{(x-\mu)^2}{\nu\sigma^2}\right)\right)$$

Let  $C = \frac{1}{\sqrt{\pi \nu \sigma^2} \Gamma(\nu/2)}$ 

$$f(x) = \int f(x, w) dw = C \int \frac{1}{2^{(\nu+1)/2}} w^{(\nu-1)/2} \exp\left(-\frac{w}{2} \left(1 + \frac{(x-\mu)^2}{\nu \sigma^2}\right)\right) dw$$

Let  $z = \frac{\nu+1}{2}$ , implying that  $z - 1 = \frac{\nu-1}{2}$ .

$$f(x) = \int \frac{1}{2^z} w^{z-1} \exp\left(-\frac{w}{2} \left(1 + \frac{(x-\mu)^2}{\nu \sigma^2}\right)\right) dw$$

And let  $t = \frac{w}{2} \left( 1 + \frac{(x-\mu)^2}{\nu\sigma^2} \right)$ , implying that  $w = 2t \left( 1 + \frac{(x-\mu)^2}{\nu\sigma^2} \right)^{-1}$  and  $dw = 2 \left( 1 + \frac{(x-\mu)^2}{\nu\sigma^2} \right)^{-1} dt$ . Note that the limits of the integral are still the same  $((0,\infty))$ .

$$f(x) = C \int \frac{1}{2^z} w^{z-1} \exp(-t) dw$$

Substituting w and  $dw = 2\left(1 + \frac{(x-\mu)^2}{\nu\sigma^2}\right)^{-1}dt$  we obtain

$$f(x) = C \int \frac{1}{2^z} \left( 2\left(1 + \frac{(x-\mu)^2}{\nu\sigma^2}\right) \right)^{-z} t^{z-1} \exp\left(-t\right) dt = C\left(1 + \frac{(x-\mu)^2}{\nu\sigma^2}\right)^{-z} \int t^{z-1} \exp\left(-t\right) dt$$

Recognizing the gamma function we finally obtain

$$f(x) = \frac{\Gamma(\frac{\nu+1}{2})}{\sqrt{\pi\nu\sigma^2}\Gamma(\nu/2)} \left(1 + \frac{(x-\mu)^2}{\nu\sigma^2}\right)^{-\frac{\nu+1}{2}}$$

Which is recognized as the density of the t-distribution.

## Maximize the complete data log-likelihood and implement this as a function

We find the derivatives of the log-likelihood with respect to  $\mu$  and  $\sigma^2$  and set them equal to zero. This results in the following estimators.

```
sample <- generator(100)
mu_hat <- mle_mu(sample$x, sample$w)
sigma2_hat <- mle_sigma2(sample$x, sample$w, mu_hat, 5)
mu_hat; sigma2_hat</pre>
```

## [1] 0.03768418

## [1] 0.7599045

## EM algorithm

Suppose we only observe the X's. We wish to use the EM algorithm to estimate the parameters  $\mu$  and  $\sigma^2$ . We denote the pair of parameters by  $\theta = (\mu, \sigma^2)$ . The E-step of the algorithm amounts to computing the expectation

$$E_{\theta'}(\log(f(x, w|\theta))|X = x) = -n\log\sigma + \int \left(\frac{n(\nu - 1)}{2}\log(w) - \sum_{i=1}^{n} \frac{w}{2}\left(1 + \frac{(x_i - \mu)^2}{\nu\sigma^2}\right)\right) f_{w|X = x}(w)dw$$

Note that  $f_{w|X=x'}(w) = \frac{f(x',w)}{f(x')} \propto f(x',w)$ . Nu mangler du bare at genkende den tæthed som en gamma tæthed...