

### Exercise 11

One of the major discoveries of Gabor was the fact, that the product of the time and frequency extend  $\Delta t \cdot \Delta \omega$  is minimized by the family of the Gabor-elementary-functions. These functions are expressed by a complex exponential function using gaussians:

$$g(\alpha, t) = \exp\left\{-\frac{\pi t^2}{\sigma^2}\right\} \cdot \exp(j\alpha t) = \exp\left\{-\frac{\pi t^2}{\sigma^2} + j\alpha t\right\}$$

The effective extend  $\Delta x$  of a function  $f(x)$  is the square-root of the variance of the power-distribution:

$$(\Delta x)^2 = \frac{\int_{-\infty}^{\infty} (x - \bar{x})^2 \cdot f(x) f^*(x) dx}{\int_{-\infty}^{\infty} f(x) f^*(x) dx}$$

where  $f^*$  denotes the conjunctive complex value of  $f$  and  $\bar{x}$  denotes the mean-value of  $x$ . Show, that for Gabor-functions the inequality

$$\Delta t \cdot \Delta \omega \geq \frac{1}{2}$$

always equals 1/2.

### Exercise 12

Compare the sampling frequency  $\Delta \omega$  of the Gabor-transform, that yields  $\Omega \leq \frac{2\pi}{\tau_0}$  where  $\tau_0$  denotes the length of the actual frame, with the discrete fragmentations of the frequencies of the DFT. For which case are both fragmentations identical?

### Exercise 13

Construct a multilayer neural network for the training procedure of the Gabor-coefficients using a perceptron. The input layer consists of the Gabor-elementary functions and the signal  $\hat{y}$ . The output layer computes all of the Gabor-coefficients.