

# Worksheet 2

Practical Lab Numerical Computing

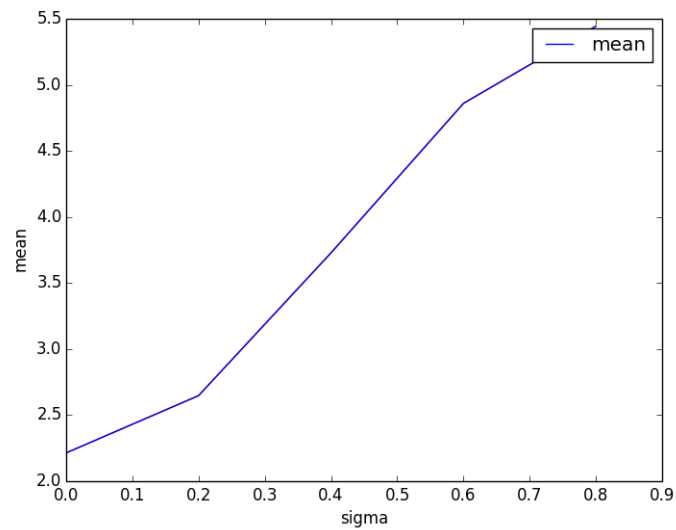
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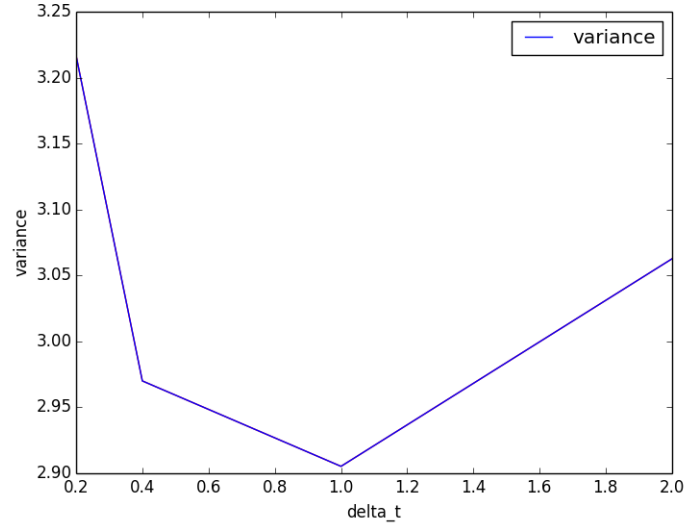
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## Task 1



## Task 2



## Task 3

To prove that

$$\mathbb{E}[V_{call}(S_T, 0)] = S(0) \exp(\mu T) \Phi(\sigma\sqrt{T} - \chi) - K \Phi(-\chi)$$

we use that by change-of-variable with  $t := -t$  we get

$$\begin{aligned}
 & \mathbb{E}[V_{call}(S_T, 0)] \\
 &= \frac{1}{\sqrt{2\pi}} \int_{\chi}^{\infty} \left( S(0) \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right) \cdot T + \sigma\sqrt{T}s\right) - K \right) \exp\left(-\frac{s^2}{2}\right) ds \\
 &= \frac{1}{2\pi} \int_{\chi}^{\infty} \left( S(0) \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right) \cdot T + \sigma\sqrt{T}s\right) \right) \exp\left(-\frac{s^2}{2}\right) ds - K \frac{1}{2\pi} \int_{\chi}^{\infty} \exp\left(-\frac{t^2}{2}\right) dt \\
 &= \Psi - K \frac{1}{2\pi} \int_{-\infty}^{-\chi} \exp\left(-\frac{t^2}{2}\right) dt \\
 &= \Psi - K \Phi(-\chi)
 \end{aligned}$$

with

$$\Psi := \frac{1}{2\pi} \int_{\chi}^{\infty} \left( S(0) \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right) \cdot T + \sigma\sqrt{T}s\right) \right) \exp\left(-\frac{s^2}{2}\right) ds.$$

Now we prove that

$$\Psi = S(0) \exp(\mu T) \Phi(\sigma\sqrt{T} - \chi).$$

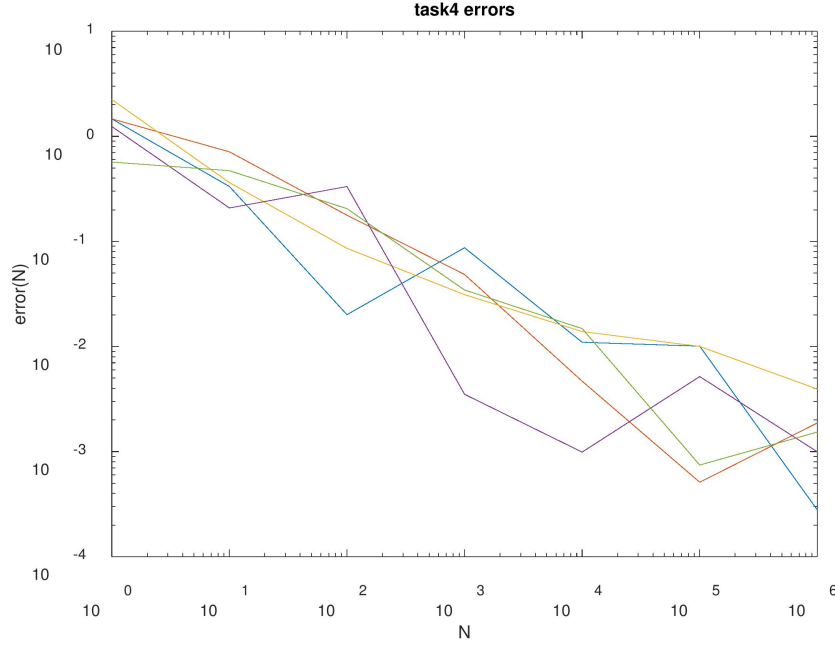
Again, we use a change-of-variable  $z := t + \sigma\sqrt{T}$  and get

$$\begin{aligned}
& \frac{1}{\sqrt{2\pi}} \exp\left(\frac{\sigma^2}{2}T\right) \int_{-\infty}^{\sigma\sqrt{T}-\chi} \exp\left(-\frac{t^2}{2}\right) dt \\
&= \frac{1}{\sqrt{2\pi}} \exp\left(\frac{\sigma^2}{2}T\right) \int_{\chi-\sigma\sqrt{T}}^{\infty} \exp\left(-\frac{t^2}{2}\right) dt \\
&= \frac{1}{\sqrt{2\pi}} \exp\left(\frac{\sigma^2}{2}T\right) \int_{\chi}^{\infty} \exp\left(-\frac{(\sigma\sqrt{T}-z)^2}{2}\right) dz \\
&= \frac{1}{\sqrt{2\pi}} \int_{\chi}^{\infty} \exp\left(-\frac{z^2}{2} + z\sigma\sqrt{T}\right) dz.
\end{aligned}$$

We have

$$\begin{aligned}
\Psi &= \frac{1}{\sqrt{2\pi}} S(0) \exp\left(\left(\mu - \frac{\sigma^2}{2}\right) \cdot T\right) \int_{\chi}^{\infty} \exp\left(-\frac{s^2}{2} + \sigma\sqrt{T}s\right) ds \\
&= S(0) \exp(\mu T) \exp\left(-\frac{\sigma^2}{2}T\right) \frac{1}{\sqrt{2\pi}} \int_{\chi}^{\infty} \exp\left(-\frac{s^2}{2} + \sigma\sqrt{T}s\right) ds \\
&= S(0) \exp(\mu T) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{\sigma^2}{2}T\right) \exp\left(\frac{\sigma^2}{2}T\right) \int_{-\infty}^{\sigma\sqrt{T}-\chi} \exp\left(-\frac{s^2}{2}\right) ds \\
&= S(0) \exp(\mu T) \Phi(\sigma\sqrt{T} - \chi).
\end{aligned}$$

## Task 4



## Task 5

We have  $\Phi^{-1}(0) = -\infty$  and  $\Phi^{-1}(1) = \infty$  where  $\Phi^{-1} : (0, 1) \rightarrow (-\infty, \infty)$  is the inverse cumulative distribution function. As an integral of a positive continuous function,  $\Phi$  is a bijection, continuous and differentiable. That means  $\Phi$  is a diffeomorphism.  $\Phi^{-1}$  is also differentiable, so we can use the transformation theorem with the change-of-variable  $t = \Phi(s)$ . We have  $|\det(D\Phi(s))| = \frac{1}{\sqrt{2\pi}} \exp(-\frac{s^2}{2})$  and

$$\begin{aligned}
 & \int_0^1 f(\Phi^{-1}(t)) dt \\
 &= \int_{\Phi^{-1}(0)}^{\Phi^{-1}(1)} f(\Phi^{-1}(\Phi(s))) \cdot |\det(D\Phi(s))| ds \\
 &= \int_{-\infty}^{\infty} f(s) \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{s^2}{2}\right) ds \\
 &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} f(s) \exp\left(-\frac{s^2}{2}\right) ds
 \end{aligned}$$

what proves formula (7).

## Task 6

In case of the trapezoidal-rule, the set of nodes of level  $l$  is a subset of the nodes of level  $l + 1$ . Furthermore, the  $2^l$  additional values lay exactly half way in between the nodes of level  $l$ . (Except for the first and the last node, which lie half way in between 0 and the first node of level  $l$  and in between the last node of level  $l$  and the end of the interval.)

## Task 7

In case of the Gauß-Legendre Quadrature, the nodes of level  $l$  are not a subset of the nodes of level  $l + 1$ . At the edges of the interval, the nodes are denser. According to Satz 1.18 from Einführung in die Grundlagen der Numerik, node  $x_i^{(N_l)}$  from level  $l$  lays between the nodes  $x_i^{(N_{l+1})}, x_{i+1}^{(N_{l+1})}$  from level  $l + 1$ .

The nodes are the roots of a three-term-recurrence relation of the form  $p_{n+1}(t) = (t - \alpha_n)p_n(t) - \beta_n^2 p_{n-1}(t)$ ,  $n \geq 0$ . The roots are the eigenvalues of a tridiagonal  $(N_l \times N_l)$ -dimensional matrix of the form

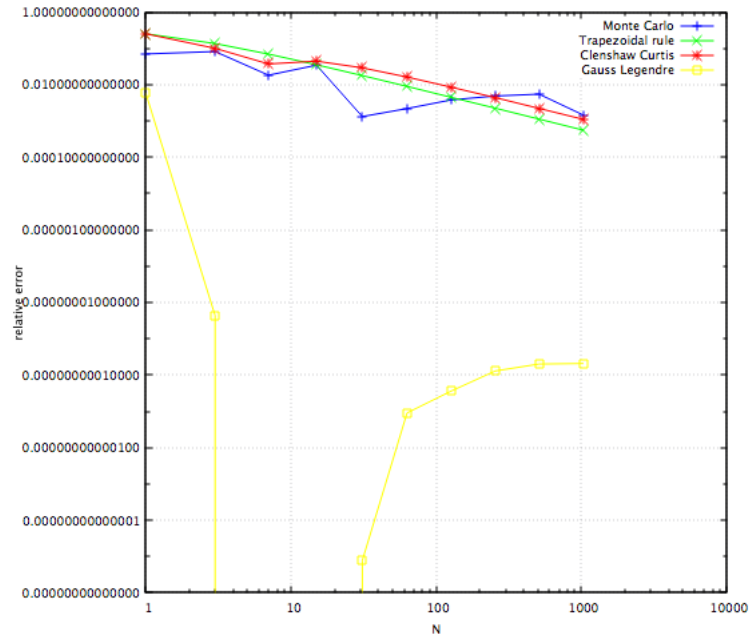
$$\begin{bmatrix} \alpha_0 & \beta_1 & 0 & \dots & \dots & \dots \\ \beta_1 & \alpha_1 & \beta_2 & \dots & \dots & \dots \\ 0 & \beta_2 & \alpha_2 & \beta_3 & \dots & \dots \\ 0 & \dots & \dots & \dots & \dots & 0 \\ \dots & \dots & 0 & \beta_{N_l-2} & \alpha_{N_l-2} & \beta_{N_l-1} \\ \dots & \dots & \dots & 0 & \beta_{N_l-1} & \alpha_{N_l-1} \end{bmatrix}$$

The weights are the first entry of the eigenvectors to the eigenvalues calculated for the nodes. Alternitavely, they can be recieved by taking  $\Lambda_{n+1}(x_i)$  of the Christoffel-function, with  $x_i, i = 1, \dots, n$  nodes.

## Task 8

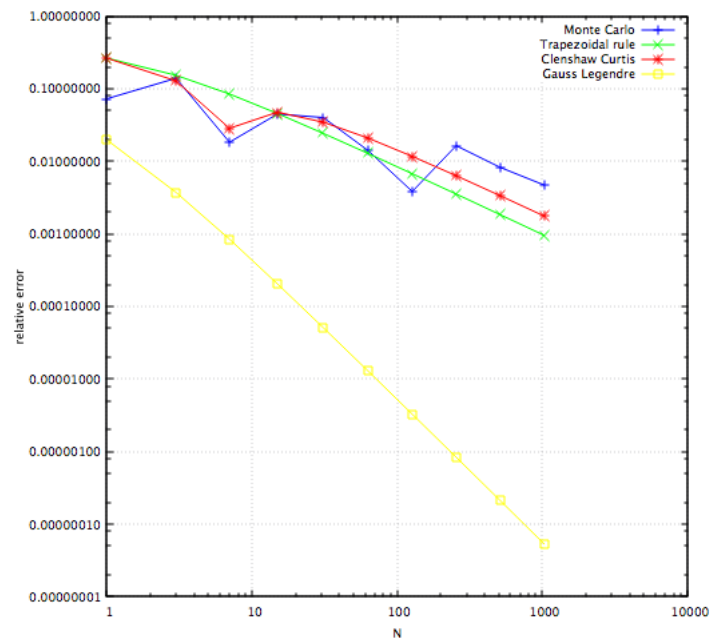
The Clenshaw-Curtis Quadrature rule uses nested nodes, the nodes of each level are a subset of the nodes of higher levels. Similarly to Gauß-Legendre quadrature, the number of nodes is higher at the edges of the interval.

## Task 9



## Task 10

For  $K = 0$ :



For  $K = 10$ :

