## Exercise 3.3

a) Let g be a strictly increasing continuously differentiable function. Then we have the Taylor series expansion:

$$g(\widehat{A}(t)) \approx g(A(t)) + g'(A(t))(\widehat{A}(t) - A(t))$$

It follows that the distribution of  $g(\widehat{A}(t))$  is approximately the same as the distribution of

$$g(A(t)) + g'(A(t))(\widehat{A}(t) - A(t))$$
(E.1)

If we use that  $\widehat{A}(t)$  is approximately normally distributed with mean A(t), we see that (E.1) is approximately normally distributed with mean g(A(t)) and variance  $\{g'(A(t))\}^2 \operatorname{Var} \widehat{A}(t)$ . The latter is estimated by  $\{g'(\widehat{A}(t))\}^2 \widehat{\sigma}^2(t)$ . This shows the result.

b) By the result in question a we have that

$$\frac{g(\widehat{A}(t)) - g(A(t))}{g'(\widehat{A}(t))\widehat{\sigma}(t)}$$

is approximately standard normally distributed. It then follows by a standard argument that

$$g(\widehat{A}(t)) \pm z_{1-\alpha/2} g'(\widehat{A}(t)) \widehat{\sigma}(t)$$

is an approximate  $100(1 - \alpha)$ % confidence interval for g(A(t)). Here  $\pm$  means + for the upper limit and – for the lower limit of the confidence interval.

c) With  $g(x) = \log x$ , we have g'(x) = 1/x. Thus an approximate  $100(1 - \alpha)\%$  confidence interval for  $\log A(t)$  is

$$\log \widehat{A}(t) \pm z_{1-\alpha/2} \,\widehat{\sigma}(t) / \widehat{A}(t)$$

By exponentiating the lower and upper limits of this confidence interval, we get the following confidence interval for A(t):

$$\exp\left\{\log\,\widehat{A}(t)\pm z_{1-\alpha/2}\,\widehat{\sigma}(t)/\widehat{A}(t)\right\}$$

i.e.

$$\widehat{A}(t) \exp\left\{\pm z_{1-\alpha/2}\,\widehat{\sigma}(t)/\widehat{A}(t)\right\}$$

This is (3.7) in the ABG-book.