# Modelling and System Identification - Microexam 1 

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Surname:
Name:
Matriculation number:

Study:

## Studiengang: Bachelor $\square$ Master $\square$

Please fill in your name above and tick exactly one box for the right answer of each question below.

1. What is the probability density function (PDF) $p_{X}(x)$ for a normally distributed random variable $X$ with mean -3 and standard deviation 3? The answer is $p_{X}(x)=\frac{1}{\sqrt{2 \pi 9}} \cdots$
(a) $\square e^{-\frac{(x+3)^{2}}{6}}$
(b) $\square \quad e^{-\frac{(x+3)^{2}}{18}}$
(c) $\square e^{\frac{(x-3)^{2}}{18}}$
(d) $\square e^{\frac{(x-3)^{2}}{6}}$
2. What does the term $\frac{1}{\sqrt{2 \pi 9}}$ in $p_{X}(x)$ ensure?
(a) $\square \quad \int_{-\infty}^{\infty} p(x)=1$
(b) $\square \quad p(x)>0$
(c) $\square \quad p(x) \geq 0$
(d) $\square$ Nothing
3. What is the PDF of a variable $y$ with uniform distribution on the interval $[5,7]$ ? For $z \in[5,7]$ it has the value:
(a) $\square \quad p_{z}(y)=\frac{1}{2^{2}}$
(b) $\square \quad p_{z}(y)=\frac{1}{2}$
(c) $\square \quad p_{y}(z)=\frac{1}{\sqrt{2}}$
(d) $\square \quad p_{y}(z)=\frac{1}{2}$
4. What is the PDF of an $n$-dimensional normally distributed variable $Z$ with zero mean and covariance matrix $\Sigma \succ 0$ ? The answer is $p_{Z}(x)=\ldots$
(a) $\square \frac{1}{\sqrt{(2 \pi)^{n} \operatorname{trace}(\Sigma)}} e^{-\frac{1}{2} x^{T} \Sigma x}$
(b) $\square \frac{1}{\sqrt{(2 \pi)^{n} \operatorname{det}(\Sigma)}} e^{-\frac{1}{2} x^{T} \Sigma^{-1} x}$
(c) $\square \frac{1}{\sqrt{2 \pi \operatorname{det}(\Sigma)}} e^{-\frac{1}{2} x^{T} \Sigma^{-1} x}$
(d) $\square \frac{1}{\sqrt{2 \pi \operatorname{trace}(\Sigma)}} e^{\frac{1}{2} x^{T} \Sigma^{-1} x}$
5. Regard a random variable $X \in \mathbb{R}^{n}$ with mean $\mu \in \mathbb{R}^{n}$ and covariance matrix $\Sigma \in \mathbb{R}^{n \times n}$. For a fixed $b \in \mathbb{R}^{m}$ and $D, A \in \mathbb{R}^{m \times n}$, regard another random variable $Y$ defined by $Y=A b+D X$. What is the covariance matrix of $Y$ ?
(a) $\square D \Sigma D^{T}$
(b) $\square \quad A^{T} \Sigma^{-1} A$
(c) $\square$
$D^{-1} \Sigma\left(D^{T}\right)^{-1}$
(d) $\square \quad D \Sigma^{-1} D^{T}$
6. Above in Question 5, what is the mean of the matrix valued random variable $Z=Y Y^{T}$ ?
(a) $\square(A b+D \mu)(A b+D \mu)^{T}+D \Sigma D^{T}$
(b) $\square(A b+D \mu)(A b+D \mu)^{T}$
(c) $\square \quad A b b^{T} A^{T}+2 A b \mu^{T} D^{T}+D \Sigma D^{T}$
(d) $\square \quad b^{T} A^{T} A b+2 \mu^{T} D^{T} A b+b^{T} \Sigma D^{T}$
7. A scalar random variable has the variance $w$. What is its standard deviation?
(a) $\square \quad w$
(b) $\square \quad w^{-1}$
(c) $\square \quad w^{2}$
(d) $\square \sqrt{w}$
8. Regard a random variable $\lambda \in \mathbb{R}$ with zero mean and standard deviation $d$. What is the mean of the random variable $y=\lambda^{2}$ ?
(a) $\square 0$
(b) $\square d$
(c) $\square d^{2}$
(d) $\square \quad \lambda+d$
9. Regard a random variable $X \in \mathbb{R}^{n}$ with zero mean and covariance matrix $\Sigma$. Given a vector $c \in \mathbb{R}^{n}$, what is the mean of $Z=c^{T} X X^{T} c$ ?
(a) $\square \operatorname{det}(\Sigma)$
(b)
$\square$
$c^{T} \operatorname{trace}(\Sigma) c$
(c) $\square \quad c^{T} \Sigma c$
(d) $\square \quad c^{T} c \operatorname{trace}(\Sigma)$
10. What is the minimizer $x^{*}$ of the convex function $f: \mathbb{R}_{++} \rightarrow \mathbb{R}, f(x)=-\log (x)+5 x$ ?
(a) $\square \quad x^{*}=-5$
(b)
$x^{*}=1 / 5$
(c) $\square \quad x^{*}=e^{5}-1$
(d) $\square \quad x^{*}=5$
11. What is the minimizer $x^{*}$ of the convex function $f: \mathbb{R} \rightarrow \mathbb{R}, f(x)=\alpha+\alpha y^{2}-\frac{1}{2} \beta y$ with $\beta>0$ ?
(a) $\square \quad x^{*}=\frac{\beta}{\alpha}$
(b) $\square$
$x^{*}=\frac{\beta}{4 \alpha}$
(c) $\square \quad x^{*}=\frac{\alpha}{\beta}$
(d) $\square \quad x^{*}=\frac{2 \beta}{\alpha}$
12. What is the minimizer of the function $f: \mathbb{R}^{n} \rightarrow \mathbb{R}, f(x)=\left\|-b+D^{T} x\right\|_{2}^{2}$ (with $D^{T}$ of rank $n$ )? The answer is $x^{*}=\ldots$
(a)
$\square-\left(D^{T} D\right)^{-1} D^{T} b$
(b) $\square$
$\left(D D^{T}\right)^{-1} D b$
(c) $\square$
$\square-\left(D D^{T}\right)^{-1} D b$
(d) $\square \quad\left(D^{T} D\right)^{-1} D^{T} b$
13. For a matrix $\Phi \in \mathbb{R}^{N \times d}$ with rank $d$, what is its pseudo-inverse $\Phi^{+}$?
(a) $\square \quad\left(\Phi \Phi^{T}\right)^{-1} \Phi^{T}$
(b) $\square \quad\left(\Phi \Phi^{T}\right)^{-1} \Phi$
(c)
(c) $\square$
$\left(\Phi^{T} \Phi\right)^{-1} \Phi^{T}$
(d) $\square \quad\left(\Phi^{T} \Phi\right)^{-1} \Phi$
14. Given a sequence of numbers $y(1), \ldots, y(N)$, what is the minimizer $\theta^{*}$ of the function $f(\theta)=\sum_{k=1}^{N}(y(k)-3 \theta)^{2}$ ?
(a) $\square \frac{1}{3 N} \sum_{k=1}^{N} y(k)^{2}$
(b)
$\square \frac{\sum_{k=1}^{N} y(k)}{3 N}$
(c) $\square \frac{1}{9 N} \sum_{k=1}^{N} y(k)^{2}$
(d) $\square \frac{\sum_{k=1}^{N} y(k)}{9 N}$
15. Given a prediction model $y(k)=\theta_{2} x(k)+2 \theta_{1}+\theta_{3} x(k)^{3}+\epsilon(k)$ with unknown parameter vector $\theta=\left(\theta_{1}, \theta_{2}, \theta_{3}\right)^{T}$, and assuming i.i.d. noise $\epsilon(k)$ with zero mean, and given a sequence of $N$ scalar input and output measurements $x(1), \ldots, x(N)$ and $y(1), \ldots, y(N)$, we want to compute the linear least squares (LLS) estimate $\hat{\theta}_{N}$ by minimizing the function $f(\theta)=$ $\left\|y_{N}-\Phi_{N} \theta\right\|_{2}^{2}$. If $y_{N}=(y(1), \ldots, y(N))^{T}$, how do we need to choose the matrix $\Phi_{N} \in \mathbb{R}^{N \times 2}$ ?

16. Which of the following is NOT a name of a probability distribution?
(a) $\square$ Uniform
(b) $\square$ Gaussian
(c)
Newton
(d) $\square$ Laplace
17. Given a random variable $X$, where $X \sim \mathcal{U}[-1,1]$, regard the following $X$-dependent random variables $Y$. For one of them $X$ and $Y$ are uncorrelated, which one?
(a) $\square \quad y=\sin (x)$
(b) $\square \quad y=\cos (x)$
(c) $\square \quad y=x^{3}$
(d) $\square \quad y=\mathrm{e}^{x}$
18. Given a set of measurements $y_{N}$ following the model $y_{N}=\Phi_{N} \theta_{0}+\epsilon$, where $\Phi_{N}$ is a regression matrix, $\theta_{0}$ a vector with true parameter values and $\epsilon(k) \sim \mathcal{N}\left(0, \sigma_{\epsilon}^{2}\right)$ the noise contribution for $k=1, \ldots, N$, we can compute the LLS estimator of the parameters $\theta$ as $\hat{\theta}_{\mathrm{LS}}$. Defining the covariance of $\hat{\theta}_{\mathrm{LS}}$ as $\Sigma_{\hat{\theta}}$, which of the following is NOT true?

| (a) $\square \quad \hat{\theta}_{\mathrm{LS}}$ is a random variable | (b) $\square \quad \hat{\theta}_{\mathrm{LS}} \sim \mathcal{N}\left(\theta_{0}, \Sigma_{\hat{\theta}}\right)$ |
| :--- | :--- |
| (c) $\square \Sigma_{\hat{\theta}}=\sigma_{\epsilon}^{2}\left(\Phi_{N}^{+^{\mathrm{T}}} \Phi_{N}^{+}\right)$ | (d) $\square \quad \hat{\theta}_{\mathrm{LS}}=\Phi_{N}^{+} y_{N}$ |

19. In the case given in the previous question, if the measurements $y_{N}$ come from a single experiment, which condition does the noise require in order to be able to compute an estimate of $\sigma_{\epsilon}^{2}$ ?
20. Imagine that the condition asked in the previous exercise is not met. We know that the noise has zero mean and covariance $\Sigma_{\epsilon_{N}}$. What would be the covariance matrix $\Sigma_{\hat{\theta}}$ of the unweighted LLS estimate?

| (a) $\square \Sigma_{\epsilon_{N}} \Phi_{N}^{+\mathrm{T}} \Phi_{N}^{+}$ | (b) $\square \Sigma_{\epsilon_{N}}^{-1} \Phi_{N}^{\mathrm{T}} \Phi_{N}$ |  |
| :--- | :--- | :--- |
| (c) $\square$ | $\Phi_{N}^{\mathrm{T}} \Sigma_{\epsilon_{N}}^{-1} \Phi_{N}$ | (d) $\square \Phi_{N}^{+} \Sigma_{\epsilon_{N}} \Phi_{N}^{+{ }^{\mathrm{T}}}$ |

