STA500 Introduction to Probability and Statistics 2, autumn 2018.

## Solution exercise set 3

## Exercises from the book:

6.40
$X=$ consumption of water is having a gamma distribution with $\alpha=2$ and $\beta=3$.

$$
\begin{aligned}
P(X>9) & =\int_{9}^{\infty} \frac{1}{3^{2} \Gamma(2)} x^{2-1} e^{-x / 3} d x=\frac{1}{9} \int_{9}^{\infty} x e^{-x / 3} d x \\
& =\frac{1}{9}\left[x(-3) e^{-x / 3}\right]_{9}^{\infty}-\frac{1}{9} \int_{9}^{\infty}-3 e^{-x / 3} d x \\
& =\frac{1}{9} 27 e^{-3}+\frac{1}{9}\left[-9 e^{-x / 3}\right]_{9}^{\infty}=3 e^{-3}+e^{-3}=4 e^{-3}=\underline{\underline{0.199}}
\end{aligned}
$$

Here integration by parts is used to solve the integral. Alternatively the integration formula found on the last page of the formula sheets could have been used.

### 6.43

a) By the general formulas for expectation and variance in the gamma distribution we get: $\mathrm{E}(X)=\alpha \beta=2 \cdot 3=\underline{\underline{6}}$ and $\operatorname{Var}(X)=\alpha \beta^{2}=2 \cdot 3^{2}=\underline{\underline{18}}$.

### 6.47

a) For the Weibull distribution with parameters $\alpha$ and $\beta$ we have that (see the paragraph about the gamma function on the formula sheets)
$\mathrm{E}(X)=\alpha^{-1 / \beta} \Gamma\left(1+\frac{1}{\beta}\right)$. I.e.

$$
\mathrm{E}(X)=0.5^{-1 / 2} \Gamma\left(1+\frac{1}{2}\right)=\sqrt{2} \cdot \Gamma\left(\frac{3}{2}\right)=\sqrt{2}\left(\frac{3}{2}-1\right) \Gamma\left(\frac{1}{2}\right)=\sqrt{\frac{\pi}{2}}=\underline{\underline{1.25}}
$$

b)

$$
\begin{gathered}
f(x)=\alpha \beta x^{\beta-1} e^{-\alpha x^{\beta}}=x e^{-\frac{1}{2} x^{2}} \quad, x \geq 0 \quad \text { which gives that: } \\
P(X>2)=\int_{2}^{\infty} x e^{-\frac{1}{2} x^{2}} d x=\left[-e^{-\frac{1}{2} x^{2}}\right]_{2}^{\infty}=e^{-\frac{1}{2} 2^{2}}=e^{-2}=\underline{\underline{0.135}}
\end{gathered}
$$

### 6.56/6.54

We use the fact that when $X$ is having a $\operatorname{lognormal}$ distribution then $\ln (X)$ is having a normal distribution and get:

$$
\begin{aligned}
P(X>270) & =1-P(X \leq 270)=1-P(\ln (X) \leq \ln (270)) \\
& =1-P\left(\frac{\ln (X)-4}{2} \leq \frac{\ln (270)-4}{2}\right) \\
& =1-P(Z \leq 0.80)=1-0.7881=\underline{\underline{0.2119}}
\end{aligned}
$$

### 6.57/6.55

For the lognormal distribution we have that $\mathrm{E}(X)=e^{\mu+\sigma^{2} / 2}$ and $\operatorname{Var}(X)=e^{2\left(\mu+\sigma^{2}\right)}-e^{2 \mu+\sigma^{2}}$ which here gives that:

$$
\begin{aligned}
\mathrm{E}(X) & =e^{4+2^{2} / 2}=e^{6}=\underline{403.4} \\
\operatorname{Var}(X) & =e^{2\left(4+2^{2}\right)}-e^{2 \cdot 4+2^{2}}=e^{16}-e^{12}=\underline{\underline{8723355.7}}
\end{aligned}
$$

### 6.58/6.56

a) Let $X$ be the number of cars arriving at the intersection during one minute. From the information in the exercise text we get that $X$ is having a Poisson distribution with expectation 5 such that:

$$
P(X>10)=1-P(X \leq 10) \stackrel{\text { table }}{=} 1-0.9863=\underline{\underline{0.0137}}
$$

b) Here we shall calculate the probability that it will take more than 2 minutes before 10 cars have appeared at the intersection. This can be calculated in two ways.
One (the simplest) way to calculate this is to define $Y$ as the number of events in the interval $[0,2]$ and calculate $P(Y<10)$ (the event that it takes more than 2 minutes until 10 cars have appeared is the same as the event that less than 10 cars appear during the 2 minutes). $Y$ is having a Poisson distribution with expectation $\lambda t=5 \cdot 2=10$ and we get:

$$
P(Y<10)=P(Y \leq 9) \stackrel{\text { table }}{=} \underline{\underline{0.458}}
$$

The other way is to define $S_{10}$ as the time which elapses until car number 10 appears at the intersection and calculate $P\left(S_{10}>2\right)$. We know that time until event number 10 in a Poisson process with $\lambda=5$ is having a gamma distribution with parameters $\alpha=10$ and $\beta=1 / 5$ (being the sum of 10 exponentially distributed variables with expectation $1 / 5$ ). We then get that:

$$
\begin{aligned}
P\left(S_{10}>2\right) & =1-P\left(S_{10} \leq 2\right)=1-\int_{0}^{2} \frac{1}{\beta^{\alpha} \Gamma(\alpha)} s^{\alpha-1} e^{-s / \beta} d s \\
& =1-\int_{0}^{2} \frac{1}{\left(\frac{1}{5}\right)^{10} \Gamma(10)} s^{10-1} e^{-5 s} d s \stackrel{u=5 s}{=} 1-\int_{0}^{10} \frac{5^{10}}{\Gamma(10)}\left(\frac{u}{5}\right)^{10-1} e^{-u} d u / 5 \\
& =1-\int_{0}^{10} \frac{u^{10-1}}{\Gamma(10)} e^{-u} d u \stackrel{t a b l e A .24}{=} 1-0.542=\underline{\underline{0.458}}
\end{aligned}
$$

### 6.59/6.57

The time between events in a Poisson process is having an exponential distribution with expectation $1 / \lambda=1 / 5$.
a)

$$
P(X>1)=1-\int_{0}^{1} 5 e^{-5 x} d x=1-\left[-e^{-5 x}\right]_{0}^{1}=1+e^{-5}-1=\underline{\underline{0.0067}}
$$

b) $\mathrm{E}(X)=\beta=1 / 5=\underline{\underline{0.2}}$

## Exercises from the note on extreme values:

## Exercise 1:

We have a parallel system made up of two independent components, where the lifetime of each component is having an exponential distribution with parameter $\lambda$.

$$
F_{X_{i}}(t)=1-\exp (-\lambda t)
$$

Let the lifetime of the system be denoted $V$. We shall find the distribution of $V$. The system is functioning as long as at least one of the components are functioning.


Figure 1: Parallel system with two components.

$$
\begin{aligned}
F_{V}(v) & =P(V<v)=P\left(\max \left(X_{1}, X_{2}\right)<v\right)=P\left(X_{1}<v \cap X_{2}<v\right) \\
& \stackrel{\text { indep. }}{=} P\left(X_{1}<v\right) \cdot P\left(X_{2}<v\right)=\left(1-e^{-\lambda v}\right)^{2} \\
& =1-2 e^{-\lambda v}+e^{-2 \lambda v} \\
f_{V}(v) & =F_{V}^{\prime}(v)=2 \lambda e^{-\lambda v}-2 \lambda e^{-2 \lambda v}=\underline{\left.\underline{2 \lambda\left(e^{-\lambda v}\right.}-e^{-2 \lambda v}\right)}, v \geq 0
\end{aligned}
$$

Expectation:

$$
\begin{aligned}
\mathrm{E}(V) & =\int_{0}^{\infty} v f_{V}(v) d v=\int_{0}^{\infty} v 2 \lambda\left(e^{-\lambda v}-e^{-2 \lambda v}\right) d v \\
& =2 \int_{0}^{\infty} v \lambda e^{-\lambda v}-\int_{0}^{\infty} v 2 \lambda e^{-2 \lambda v}=2 \frac{1}{\lambda}-\frac{1}{2 \lambda}=\frac{3}{\underline{2 \lambda}}
\end{aligned}
$$

Notice that we recognize the two last integrals as the expression for the expectation for exponentially distributed variables with respectively parameter $\lambda$ and parameter $2 \lambda$, and we thus know what these integrals are.

## Exercise 2:

We have a series system made up of $n$ independent components where the lifetime of component $i$ is having a Weibull distribution with parameters $\alpha$ and $\beta$ :

$$
P\left(X_{i} \leq x\right)=F_{X_{i}}(x)=1-e^{-\alpha x^{\beta}} \quad, x \geq 0
$$

Let the lifetime of the system be denoted $U$. We shall first find the distribution of $U$. The system is only functioning as long as all components are functioning.

$$
\begin{gathered}
F_{U}(u)=P(U<u)=P\left(\min \left(X_{1}, X_{2}, \ldots, X_{n}\right)<u\right)=1-P\left(\min \left(X_{1}, X_{2}, \ldots, X_{n}\right)>u\right) \\
\stackrel{\text { indep. }}{=} 1-\prod_{i=1}^{n} P\left(X_{i}>u\right)=1-\prod_{i=1}^{n} e^{-\alpha u^{\beta}}=\underline{\underline{1-e^{-n \alpha u^{\beta}}}}
\end{gathered}
$$



Figure 2: Series system with $n$ components.

We see that this is a Weibull distribution with parameters $n \alpha$ and $\beta$.
For the Weibull distribution with parameters $\alpha$ and $\beta$ we have that $\mathrm{E}(X)=\alpha^{-1 / \beta} \Gamma\left(1+\frac{1}{\beta}\right)$. I.e. for a single component we have that:

$$
\mathrm{E}(X)=0.1^{-1 / 0.5} \Gamma\left(1+\frac{1}{0.5}\right)=100 \cdot \Gamma(3)=100 \cdot 2!=\underline{\underline{200}}
$$

For the whole system we get:

$$
\mathrm{E}(U)=(n \cdot 0.1)^{-1 / 0.5} \Gamma\left(1+\frac{1}{0.5}\right)=\frac{1}{n^{2}} 100 \cdot \Gamma(3)=\underline{\underline{\frac{200}{n^{2}}}}
$$

## Exercise 1:

a)

$$
\begin{aligned}
\mathrm{E}(X) & =\int_{-\infty}^{\infty} x f(x) d x=\int_{0}^{\infty} x \frac{1}{\beta} e^{-x / \beta} d x=\left[-x e^{-x / \beta}\right]_{0}^{\infty}-\int_{0}^{\infty}-e^{-x / \beta} d x=0-\left[\beta e^{-y / \beta}\right]_{0}^{\infty}=\underline{\underline{\beta}} \\
\mathrm{E}\left(X^{2}\right) & =\int_{0}^{\infty} x^{2} \frac{1}{\beta} e^{-x / \beta} d x=\left[-x^{2} e^{-x / \beta}\right]_{0}^{\infty}-\int_{0}^{\infty} 2 x\left(-e^{-x / \beta}\right) d x \\
& =0+2 \beta \int_{0}^{\infty} x \frac{1}{\beta} e^{-x / \beta} d x=2 \beta^{2} \\
\Rightarrow \operatorname{Var}(X) & =\mathrm{E}\left(X^{2}\right)-\mathrm{E}(X)^{2}=2 \beta^{2}-\beta^{2}=\underline{\underline{\beta^{2}}}
\end{aligned}
$$

Instead of integration by parts integration formulas found on the last page of the formula sheets could have been used to solve the integrals.
b) With $\beta=1000$ we get

$$
P(X>1000)=\int_{1000}^{\infty} \frac{1}{1000} e^{-x / 1000} d x=\left[-e^{-x / 1000}\right]_{1000}^{\infty}=e^{-1}=\underline{\underline{0.368}}
$$

## Exercise 2:

a) The number of telephone calls per hour is having a Poisson distribution with expectation $\mu=\lambda t=6 \cdot 1=6$, i.e.

$$
P(X>6)=1-P(X \leq 6) \stackrel{\text { table }}{=} 1-0.6063=\underline{\underline{0.3937}}
$$

b) The time until the first event in a Poisson process is having an exponential distribution with expectation $1 / \lambda=1 / 6$. Also note that 10 minutes is $1 / 6$ hour. E.g.

$$
P(T<1 / 6)=\int_{0}^{1 / 6} 6 e^{-6 x} d x=\left[-e^{-6 x}\right]_{0}^{1 / 6}=-e^{-1}+1=\underline{\underline{0.6321}}
$$

(Alternatively the exercise can be solved by looking at the probability of having at least one event in a 10 minute interval)
c) The time until the second event in a Poisson process is having a gamma distribution with parameters $\alpha=2$ and $\beta=1 / \lambda=1 / 6$. Since 20 minutes is $1 / 3$ hour we then get:

$$
\begin{aligned}
P\left(S_{2}<1 / 3\right) & =\int_{0}^{1 / 3} \frac{1}{(1 / 6)^{2} \Gamma(2)} x^{2-1} e^{-x /(1 / 6)} d x=\int_{0}^{1 / 3} 36 x e^{-6 x} d x \\
& =\left[-6 x e^{-6 x}\right]_{0}^{1 / 3}-\int_{0}^{1 / 3}\left(-6 e^{-6 x}\right) d x \\
& =-2 e^{-2}-\left[e^{-6 x}\right]_{0}^{1 / 3}=-2 e^{-2}-e^{-2}+1=\underline{\underline{0.5940}}
\end{aligned}
$$

Alternatively we can solve the problem by defining $Y=$ the number of telephone calls in $[0,1 / 3], Y$ is then having a Poisson distribution with expectation $\mu=\lambda t=6 \cdot(1 / 3)=2$. If it takes less than 20 minutes until the second telephone call, this means that there will be at least 2 calls during the 20 first minutes ( $1 / 3$ hour) and we get:

$$
P\left(S_{2}<1 / 3\right)=P(Y \geq 2)=1-P(Y \leq 1) \stackrel{\text { table }}{=} 1-0.4060=\underline{\underline{0.5940}}
$$

d) The number of telephone calls during 7.5 hours is having a Poisson distribution with expectation $\mu=\lambda t=6 \cdot 7.5=45$. In principle this can be used to calculate $P(X>50)$ exactly, however, in this case it is easier to use the approximation to the normal distribution. Since $\mu>15$ the approximation to the normal distribution is good, and we get

$$
\begin{aligned}
P(X>50) & =1-P(X \leq 50)=1-P\left(\frac{X-45}{\sqrt{45}} \leq \frac{50+0.5-45}{\sqrt{45}}\right) \\
& =1-P(Z \leq 0.82)=1-0.7939=\underline{\underline{0.2061}}
\end{aligned}
$$

e) The number of events in any interval of length $t$ in a Poisson process is having a Poisson distribution with expectation $\lambda t$, i.e. the number of telephone calls during 10 minutes $=$ $1 / 6$ hour is having a Poisson distribution with expectation $\mu=6 \cdot(1 / 6)=1$, i.e.

$$
P(X=0)=\frac{1^{0}}{0!} e^{-1}=\underline{\underline{0.3679}}
$$

f) Since the number of events in non-overlapping intervals in a Poisson process are indendent (independent increments) what has happened in the previous 10 minutes has no influence on what will happen the next 10 minutes - i.e. the probability of no telephone calls the next 10 minutes is the same as the probability of no calls in any 10 minute interval, $\underline{\underline{0.3679}}$.

## Exercise 3:

a) We first find:

$$
F_{X}(x)=\int_{0}^{x} 0.02 u e^{-0.01 u^{2}} d u=\left[-e^{-0.01 u^{2}}\right]=1-e^{-0.01 x^{2}}
$$

Further we have:

$$
\begin{aligned}
F_{U}(u) & =P(U<u)=P\left(\min \left(X_{1}, X_{2}, X_{3}\right)<u\right)=1-P\left(\min \left(X_{1}, X_{2}, X_{3}\right)>u\right) \\
& \stackrel{\text { indep. }}{=} 1-P\left(X_{1}>u\right) \cdot P\left(X_{2}>u\right) \cdot P\left(X_{3}>u\right)=1-\left[1-F_{X}(u)\right]^{3} \\
& =1-\left[e^{-0.01 u^{2}}\right]^{3}=\underline{\underline{1-e^{-0.03 u^{2}}}}
\end{aligned}
$$

We could also calculate the probability density: $\quad f_{U}(u)=F_{U}^{\prime}(u)=\underline{\underline{0.06 u e^{-0.03 u^{2}}}}, u>0$
b)

$$
\begin{aligned}
& P(X<5)=F_{X}(5)=1-e^{-0.01 \cdot 5^{2}}=\underline{\underline{0.221}} \\
& P(U<5)=F_{U}(5)=1-e^{-0.03 \cdot 5^{2}}=\underline{\underline{0.528}}
\end{aligned}
$$

c) If we compare the probability density in a) with the Weibull distribution,

$$
f(x)=\alpha \beta x^{\beta-1} e^{-\alpha x^{\beta}} \quad, x \geq 0
$$

we see that $X$ is having a Weibull distribution with $\alpha=0.01$ and $\beta=2$, while $U$ is having a Weibull distribution with $\alpha=0.03$ and $\beta=2$. From the expression for the expectation in the Weibull distribution, $\mathrm{E}(X)=\alpha^{-1 / \beta} \Gamma\left(1+\frac{1}{\beta}\right)$ we then get

$$
\begin{aligned}
& \mathrm{E}(X)=0.01^{-1 / 2} \Gamma\left(1+\frac{1}{2}\right)=0.01^{-1 / 2} \frac{1}{2} \Gamma\left(\frac{1}{2}\right)=0.01^{-1 / 2} \frac{1}{2} \sqrt{\pi}=\underline{\underline{8.86}} \\
& \mathrm{E}(U)=0.03^{-1 / 2} \Gamma\left(1+\frac{1}{2}\right)=0.03^{-1 / 2} \frac{1}{2} \Gamma\left(\frac{1}{2}\right)=0.03^{-1 / 2} \frac{1}{2} \sqrt{\pi}=\underline{\underline{5.12}}
\end{aligned}
$$

