Department of Mathematics and Statistics Stochastic processes on domains Suggestions to excercise problem sheet 2

Note. In the Problems 1-10 the j,k and n are always integers.

1. Let $(P_{t,x})$ be as in Lemma 3.14 in Lecture notes (page 23). Let

$$\mu_{(t_1,\dots,t_n)}^x(A_1 \times \dots \times A_n) = \int_{A_1} P_{t_1,x}(\,\mathrm{d} x_1) \dots \int_{A_n} P_{t_n-t_{n-1},x_{n-1}}(\,\mathrm{d} x_n)$$

for every $x \in S$, for every $n \in \mathbb{N}_+$, for every $t_1 < \cdots < t_n$ and for every $A_1, \ldots, A_n \in \mathscr{S}$. Let $(A_n) \subset \mathscr{S}$ be a sequence of sets in \mathscr{S} . Define $B_{k,j}, C_{k,j} \in \mathscr{S}$ as $[B_{k,j}] = [A_k][k \neq j] + [S][k = j]$. and $[C_{k,j}] = [A_k][k < j] + [A_{k-1}][k > j]$. Let $\pi_{j,n}(t_1, \ldots, t_n) = (s_1, \ldots, s_{n-1})$ where $s_k = t_k[k < j] + t_{k-1}[k > j]$. Show that the measures $\mu_{(t_1,\ldots,t_n)}^x$ are consistent, i.e.

i) show that

$$\mu_{(t_1,\ldots,t_n)}^x(B_{1,j}\times\cdots\times B_{n,j})=\mu_{\pi_{j,n}(t_1,\ldots,t_n)}^x(C_{1,j}\times\cdots\times C_{n-1,j})$$

for every $n \in \mathbb{N}$, for every $1 \leq j \leq n$.

Suggestion. There is an unfortunate misprint s_k should be $s_k = t_k [k < j] + t_{k+1} [k \ge j]$ and similarly $C_{k,j}$ should have $[A_{k+1}][k \ge j]$. Furthermore, the *n* should be at least 2 for this to be meaningful.

Therefore, this problem is considered to solved by everyone.

With this change let's look for different values of n. First is n = 2. Then we have two equations

$$\mu_{(t_1,t_2)}^x(A_1 \times S) = \mu_{(t_1)}^x(A_1).$$

and

$$\mu_{(t_1,t_2)}^x(S \times A_2) = \mu_{(t_2)}^x(A_2).$$

Therefore, by definition these reduce to equations

$$\int_{A_1} P_{t_2 - t_1, x_1}(S) P_{t_1, x}(dx_1) = P_{t_1, x}(A_1)$$

and

$$\int_{S} P_{t_2-t_1,x_1}(A_2) P_{t_1,x}(dx_1) = P_{t_2,x}(A_2)$$

The first one follows since $P_{t,x}$ is a probability measure for every t and x and hence $P_{t_2-t_1,x_1}(S) = 1$ and thus

$$\int_{A_1} P_{t_2-t_1,x_1}(S) P_{t_1,x}(dx_1) = \int_{A_1} P_{t_1,x}(dx_1) = P_{t_1,x}(A_1).$$

The second one is the Chapman–Kolmogorov equation (since $t_2 - t_1 + t_1 = t_2$) so it holds as well. Let's suppose that the claim holds for all n (say less than N) and let's consider the case n + 1. So we have n + 1 equations to show. Suppose k = n + 1first. Then

$$\mu_{(t_1,t_2,\dots,t_{n+1})}^{(x)}(A_1 \times A_2 \times \dots \times S) = \int_{A_1} \dots \int_{A_n} P_{t_1,x}(dx_1) \dots P_{t'_n,x_{n-1}}(A_n) P_{t'_{n+1},x_n}(S)$$
$$= \mu_{(t_1,\dots,t_n)}^{(x)}(A_1 \times \dots \times A_n)$$

as in the case when n = 2. Next suppose $1 < k \leq n$. Since the equation is then of form

$$\int_{A_1} \dots \int_{A_{k-1}} P_{t'_{k-1}}(\,\mathrm{d} x_{k-1}) f(x_{k-1}) = \int_{A_1} \dots \int_{A_{k-1}} P_{t'_{k-1}}(\,\mathrm{d} x_{k-1}) g(x_{k-1})$$

it is enough to show that f = g. Now we notice that

$$f(x_{k-1}) = \mu_{(t'_k, v_{k+1}, t'_{k+2}, \dots, t'_{n+1})}^{(x_{k-1})} (S \times A_{k+1} \times \dots \times A_{n+1})$$

and

$$g(x_{k-1}) = \mu_{(v_{k+1}, t'_{k+2}, \dots, t'_{n+1})}^{(x_{k-1})} (A_{k+1} \times \dots \times A_{n+1})$$

where $v_{k+1} = t_{k+1} - t_{k-1}$. Since now there are at most n + 1 - 1 times and sets left, we get that f = g by induction assumption.

So we are left with the equation for k = 1. This time we notice that the equation is of form

$$\int_{S} P_{t_1,x}(\mathrm{d}x_1) \int_{A_2} P_{t_2',x}(\mathrm{d}x_2) f(x_2) = \int_{A_2} P_{t_2,x}(\mathrm{d}x_2) f(x_2)$$

Notice that if f would be a constant function, this would be the Chapman–Kolmogorov equation. Now Chapman–Kolmogorov equation implies for simple functions f that

$$\int_{S} P_{t_1,x}(\mathrm{d}x_1) \int_{A_2} P_{t_2',x}(\mathrm{d}x_2) f(x_2) = \int_{A_2} P_{t_2,x}(\mathrm{d}x_2) f(x_2)$$

holds. And from this the claim follows by the monotone convergence.

2. Show that the transition probability operator of Brownian motion $P_t^{(B)}$ (see pages 21–22) satisfies

$$\lim_{t \to 0} P_t^{(B)} f(x) = f(x)$$

for every $x \in \mathbb{R}^d$ and every bounded and continuous f. (Hint: change of variables so that t appears only in the argument of f and dominated convergence). Suggestion. By definition

$$P_t^{(B)} f(x) = (2\pi t)^{-d/2} \int_{\mathbb{R}^d} f(y) e^{-\frac{1}{2}|x-y|^2/t} \, \mathrm{d}y.$$

By change of variable $z = (y - x)/\sqrt{t}$ we see that $f(y) = f(x + z\sqrt{t})$, the exponent $|x - y|^2/t = |z|^2$ and the $dz = t^{-d/2} dy$. So

$$P_t^{(B)} f(x) = (2\pi)^{-d/2} \int_{\mathbb{R}^d} f(x + z\sqrt{t}) e^{-\frac{1}{2}|z|^2} \, \mathrm{d}z.$$

Since f is bounded, and $e^{-\frac{1}{2}|z|^2}$ is integrable, we have by dominated convergence that

$$\lim_{t \to 0} P_t^{(B)} f(x) = (2\pi)^{-d/2} \int_{\mathbb{R}^d} \lim_{t \to 0} f(x + z\sqrt{t}) e^{-\frac{1}{2}|z|^2} dz$$
$$= f(x)(2\pi)^{-d/2} \int_{\mathbb{R}^d} e^{-\frac{1}{2}|z|^2} dz.$$

Now the integral on the right is the Gaussian integral and so the right-hand side is 1 and the claim follows.

3. Show that the transition probability operator of Brownian motion $P_t^{(B)}$ (see 2. above) satisfies

$$\lim_{t \to 0} \|P_t^{(B)} f - f\| = \lim_{t \to 0} \sup_{x \in \mathbb{R}^d} |P_t^{(B)} f(x) - f(x)| = 0$$

for every $x \in \mathbb{R}^d$ and every $f \in C_{\infty}(\mathbb{R}^d)$. (Hint: show that there is large R such that outside ball of radius R the difference is small and 2.)

Suggestion. Let $\varepsilon > 0$. Choose $R_0 > 0$ so that $|f(x)| < \varepsilon$ for every $|x| > R_0$. So let's take any point x, such that $|x| > R > R_0$ and let's divide the integration in to two parts

$$|P_t^{(B)}f(x)| \le (2\pi t)^{-d/2} \int_{|y| \le r} \mathrm{d}y \dots + (2\pi t)^{-d/2} \int_{|y| > r} \mathrm{d}y \dots$$

Since f is bounded, say $|f(y)| \leq M$ for every $y \in \mathbb{R}^d$, we can estimate the first one by using polar coordinates to be at most

$$M(2\pi t)^{-d/2} \int_0^r e^{-(\lambda - |x|)^2/2} \lambda^{d-1} \,\mathrm{d}\lambda.$$

If r < R/2, say, then $|\lambda - |x|| = |x| - \lambda > R/2$ and hence $-(\lambda - |x|)^2 \le -R^2/4$. So we may estimate the first integral by

$$M'e^{-R^2/32}r^d \le M''e^{-R^2/8}R^d$$

Now the second integral is easy to estimate, if we assume that $r > R_0$, since then $|f(y)| < \varepsilon$ for every |y| > r and the remainding part is just the Gaussian integral. So if we can choose r < R/2 and $r > R_0$, then

$$|P_t^{(B)}f(x)| \lesssim \varepsilon + e^{-R^2/8}R^d.$$

This means that the estimate can be done provided that $R_0 < r < R/2$ which means that the estimate holds whenever $R > 2R_0$. Moreover, we can make this estimate smaller than 2ε for large enough R. Let's call such an R as R_1 .

Now for R_1 we have

$$|P_t^{(B)}f(x) - f(x)| \le 2\varepsilon + |f(x)| \le 3\varepsilon$$

for every $|x| > R_1$. Since the set $|x| \le R_1$ is bounded and closed, i.e. compact, we know that the pointwise convergence implies uniform convergence so that

$$\lim_{t \to 0} \sup_{|x| \le R_1} |P_t^{(B)} f(x) - f(x)| = 0$$

if

$$\lim_{t \to 0} |P_t^{(B)} f(x) - f(x)| = 0$$

for every $|x| \leq R_1$. But this we know from the Problem 2. Therefore, if we choose $t_0 > 0$ small enough, we have that

$$\sup_{|x| \le R_1} |P_t^{(B)} f(x) - f(x)| \le 3\varepsilon$$

for every $t < t_0$ and so

$$\sup_{x \in \mathbb{R}^d} |P_t^{(B)} f(x) - f(x)| \le 3\varepsilon$$

for every $t < t_0$. This implies the claim.

Some theory. Suppose (X_t) is a Markov process and suppose there exists a *shift* operators $\theta_s \colon \Omega \to \Omega$ for every s > 0 with the property $X_t \circ \theta_s(\omega) = X_t(\theta_s(\omega)) := X_{t+s}(\omega)$ for every $t \ge 0$ and for every $\omega \in \Omega$. This way we can easily express that general idea of forgetting the past and the restarting the clock.

Note. In Problems 4, 5, 6 and 7 the triplet $(X_t, \mathscr{F}_t, \{\mathbf{P}_x\})$ is a Markov process. The algebra \mathscr{A}_{∞} and the σ -algebre \mathscr{F}_{∞} are defined in the proof of Lemma 3.14 on page 24. The Dynkin system is defined in the proof of Lemma 3.14 on page 25.

4. Let $s < t_1 < \cdots < t_n$. Show that the time stationary Markov property for (X_t) with respect to (\mathscr{F}_t) implies that

$$\mathbf{E}_x(f_1(X_{t_1})\dots f_n(X_{t_n}) | \mathscr{F}_s) = \mathbf{E}_{X_s}(f_1(X_{t_1-s})\dots f_n(X_{t_n-s}))$$

for every $x \in S$ and for every f_j bounded and measurable function with j = 1, ..., n. You can assume that this is know for n = 1.

Suggestion. When n = 1 this claim states that

$$\mathbf{E}_x(f_1(X_{t_1}) \,|\, \mathscr{F}_s) = \mathbf{E}_{X_s} f_1(X_{t_1-s})$$

This is the Markov property in the integrated form since suppose f_1 is a simple function $a_1[B_1] + \cdots + a_n[B_n]$. Then

$$\mathbf{E}_{x}(f_{1}(X_{t_{1}}) | \mathscr{F}_{s}) = \sum_{k} a_{k} \mathbf{E}_{x}([X_{t_{1}} \in B_{k}] | \mathscr{F}_{s}) = \sum_{k} a_{k} \mathbf{E}_{X_{s}} [X_{t_{1}-s} \in B_{k}]$$
$$= \mathbf{E}_{X_{s}} f_{1}(X_{t_{1}-s})$$

Therefore, by the usual procedure we can use monotone convergence to show this for every bounded and measurable function f_1 . But this part was something you could assume to be known.

Assume that the claim holds for $N \leq n$ and let's consider the case N = n + 1. Since $f_1(X_{t_1})$ is \mathscr{F}_{t_1} -measurables, we have by the property of the conditional expectation that

$$\mathbf{E}_x(f_1(X_{t_1})\dots f_n(X_{t_n}) | \mathscr{F}_s) = \mathbf{E}_x(f_1(X_{t_1})\mathbf{E}_x(f_2(X_{t_2})\dots f_n(X_{t_n}) | \mathscr{F}_{t_1}) | \mathscr{F}_s) .$$

If you don't remember this I'll verify this in the end. But now we may use the induction assumption and we have that

$$\mathbf{E}_{x}(f_{2}(X_{t_{2}})\dots f_{n}(X_{t_{n}}) | \mathscr{F}_{t_{1}}) = \mathbf{E}_{X_{t_{1}}} f_{2}(X_{t_{2}-t_{1}})\dots f_{n}(X_{t_{n}-t_{1}}) = g(X_{t_{1}}).$$

for every x. So we have obtained that

$$\mathbf{E}_{x}(f_{1}(X_{t_{1}})\dots f_{n}(X_{t_{n}}) | \mathscr{F}_{s}) = \mathbf{E}_{x}(f_{1}(X_{t_{1}})g(X_{t_{1}}) | \mathscr{F}_{s}) = \mathbf{E}_{X_{s}}f_{1}(X_{t_{1}-s})g(X_{t_{1}-s})$$

where the last identity was shown first (or assumed to be known). Now recall

$$g(x) = \mathbf{E}_x f_2(X_{t_2-t_1}) \dots f_n(X_{t_n-t_1})$$

and so again by the induction assumption

$$g(X_{t_1-s}) = \mathbf{E}_{X_{t_1-s}} f_2(X_{t_2-t_1}) \dots f_n(X_{t_n-t_1}) = \mathbf{E}_x(f_2(X_{t_2-s}) \dots f_n(X_{t_n-s}) | \mathscr{F}_{t_1-s})$$

for every x. Especially,

$$\mathbf{E}_{x}(f_{1}(X_{t_{1}})\dots f_{n}(X_{t_{n}}) | \mathscr{F}_{s}) = \mathbf{E}_{X_{s}} f_{1}(X_{t_{1}-s}) \mathbf{E}_{X_{s}}(f_{2}(X_{t_{2}-s})\dots f_{n}(X_{t_{n}-s}) | \mathscr{F}_{t_{1}-s})$$

= $\mathbf{E}_{X_{s}} f_{1}(X_{t_{1}-s}) f_{2}(X_{t_{2}-s})\dots f_{n}(X_{t_{n}-s})$

where the latter identity follows again from the fact that $f_1(X_{t_1-s})$ is \mathscr{F}_{t_1-s} -measurable by the same property of conditional expectation we used once already.

We were using two properties of conditional expectation. One is $\mathbf{E} (XY | \mathscr{G}) = X\mathbf{E} (Y | \mathscr{G})$ when X is \mathscr{G} -measurable and the integration makes sense. This was one of the Problems in the Excercise Sheet 1. And the second is $\mathbf{E} (X | \mathscr{G}) = \mathbf{E} (\mathbf{E} (X | \mathscr{H}) | \mathscr{G})$ for every sub- σ -algebra $\mathscr{G} \subset \mathscr{H}$. This follows since this is equivalent with

$$\mathbf{E}[A]X = \mathbf{E}[A]\mathbf{E}(X|\mathscr{H})$$

for every $A \in \mathscr{G}$. When $A \in \mathscr{G} \subset \mathscr{H}$, we see that this holds by the definition of the conditional expectation with respect to \mathscr{H} .

5. Let $0 < t_1 < \cdots < t_n$ and let $A = \{X_{t_1} \in A_1, \ldots, X_{t_n} \in A_n\} \in \mathscr{A}_{\infty}$. Show that the time stationary Markov property for (X_t) with respect to (\mathscr{F}_t) implies that

$$\mathbf{E}_{x}([A] \circ \theta_{s} | \mathscr{F}_{s}) = \mathbf{E}_{X_{s}}[A]$$

$$(0.1)$$

for every $x \in S$. (Hint: try writing $[A] \circ \theta_s$ as a form of Problem 4. from the above definition of the shift operator)

Suggestion. Note that

$$[A](\omega) = [X_{t_1}(\omega) \in A_1, \dots, X_{t_n}(\omega) \in A_n]$$

and that

$$[A] \circ \theta_s(\omega) = [X_{t_1}(\theta_s(\omega)) \in A_1, \dots, X_{t_n}(\theta_s(\omega)) \in A_n]$$
$$= [X_{t_1+s}(\omega) \in A_1, \dots, X_{t_n+s}(\omega) \in A_n]$$

This means that the claim is equivalent with

$$\mathbf{P}_{x}\left(X_{t_{1}+s}\in A_{1},\ldots X_{t_{n}+s}\in A_{n}\,|\,\mathscr{F}_{s}\,\right)=\mathbf{P}_{X_{s}}\left(X_{t_{1}}\in A_{1},\ldots X_{t_{n}}\in A_{n}\,\right)$$

If we denote $t'_j = t_j + s$ and denote $f_j = [A_j]$ the claim becomes

$$\mathbf{E}_x\Big(f_1(X_{t'_1})\dots f_n(X_{t'_n})\,|\,\mathscr{F}_s\Big)\,=\,\mathbf{E}_{X_s}\,(f_1(X_{t'_1-s})\dots f_n(X_{t'_n-s}))$$

which holds by Problem 4.

6. Let \mathscr{G} be the family of sets

 $\mathscr{G} = \{ A \in \mathscr{F}_{\infty} : A \text{ satisfies the identity } (0.1) \}$

Show that \mathscr{G} is a Dynkin system and that it coincides with \mathscr{F}_{∞} . (Hint: see proof of Lemma 3.14).

Suggestion. We need to verify that

i)
$$\Omega \in \mathscr{G}$$
,

- *ii*) $A^C \in \mathscr{G}$, if $A \in \mathscr{G}$,
- *ii*) $\bigcup_k A_k \in \mathscr{G}$, if $\forall : A_k \in \mathscr{G}$ and $A_k \cap A_j = \emptyset$ for $j \neq k$.

The first one follows, since $\Omega \in \mathscr{A}_{\infty}$ and we proved that in Problem 5. The second claim is that

$$\mathbf{E}_{x}\left(\left[A^{C}\right] \circ \theta_{s} \,\middle|\, \mathscr{F}_{s}\right) = \mathbf{E}_{X_{s}}\left[A^{C}\right].$$

if $A \in \mathscr{G}$. Since $[A^C] = 1 - [A]$ the claim becomes

$$1 - \mathbf{E}_{x}([A] \circ \theta_{s} | \mathscr{F}_{s}) = 1 - \mathbf{E}_{X_{s}}[A]$$

which is equivalent with $A \in \mathscr{G}$.

The third case is similar, since for disjoint sequence of sets

$$\left[\bigcup_{k} A_{k}\right] = \sum_{k} \left[A_{k}\right]$$

so the claim of iii) becomes

$$\sum_{k} \mathbf{E}_{x}([A_{k}] \circ \theta_{s} \,|\, \mathscr{F}_{s}) \,= \sum_{k} \mathbf{E}_{X_{s}} \,[A_{k}]$$

when $(A_k) \subset \mathscr{G}$ is sequence of disjoint sets. Since $A_k \in \mathscr{G}$ means that

$$\mathbf{E}_{x}([A_{k}] \circ \theta_{s} | \mathscr{F}_{s}) = \mathbf{E}_{X_{s}}[A_{k}]$$

the claim follows. Therefore, \mathscr{G} is a Dynkin system. By the proof of Lemma 3.14. we know that $\sigma(\mathscr{A}_{\infty}) = \mathscr{F}_{\infty}$ and by Dynkin's π - λ Theorem, $\mathscr{G} \supset \sigma(\mathscr{A}_{\infty}) = \mathscr{F}_{\infty}$. But since $\mathscr{G} \subset \mathscr{F}_{\infty}$, we have that $\mathscr{G} = \mathscr{F}_{\infty}$ and the claim follows.

7. Let Z be a \mathscr{F}_{∞} -measurable bounded random variable. Show that the time stationary Markov property for (X_t) with respect to (\mathscr{F}_t) implies that

$$\mathbf{E}_x(Z \circ \theta_s \,|\, \mathscr{F}_s) \,= \mathbf{E}_{X_s} \, Z$$

for every $s \ge 0$. (Hint: first simple Z, then the general case). **Suggestion**. When $Z = a_1[A_1] + \ldots a_n[A_n]$ for $A_j \in \mathscr{F}_{\infty}$, we know by Problem 6. that $A_j \in \mathscr{G}$ and therefore,

$$\mathbf{E}_{x}([A_{k}] \circ \theta_{s} | \mathscr{F}_{s}) = \mathbf{E}_{X_{s}}[A_{k}]$$

This implies that

$$\mathbf{E}_{x}(Z \circ \theta_{s} | \mathscr{F}_{s}) = \sum_{k} a_{k} \mathbf{E}_{x}([A_{k}] \circ \theta_{s} | \mathscr{F}_{s}) = \sum_{k} a_{k} \mathbf{E}_{X_{s}}[A_{k}] = \mathbf{E}_{X_{s}} Z$$

The general case follows by monotone convergence.

8. Suppose Z is an integrable real valued random variable and (\mathscr{F}_t) is a filtration. Show that a process (X_t) which is *defined* as

$$X_t := \mathbf{E} \left(Z \,|\, \mathscr{F}_t \right)$$

is a martingale with respect to the filtration (\mathscr{F}_t) .

Suggestion. We first need to check that X is adapted, $\mathbf{E} \mid |X_t < \infty$ for every t and then the martingale property.

Since X_t is conditional expectation with respect to the filtration (\mathscr{F}_t) , it is by the definition of conditional expectation measurable \mathscr{F}_t -measurable. Thus, X is adapted.

Next, since Z is integrable, we can write $|\mathbf{E}(Z|\mathscr{F}_t)| \leq \mathbf{E}(|Z||\mathscr{F}_t)$ and therefore,

 $\mathbf{E} |X_t| \leq \mathbf{E} \mathbf{E} (|Z| | \mathscr{F}_t) = \mathbf{E} |Z| < \infty,$

and hence X_t is integrable for every t as well.

The martingale property follows from the property we have used couple of times (see Problem 4.). By definition of X_t we have

$$\mathbf{E} (X_t | \mathscr{F}_s) = \mathbf{E} (\mathbf{E} (Z | \mathscr{F}_t) | \mathscr{F}_s) .$$

Now since $\mathscr{F}_s \subset \mathscr{F}_t$, we have that

$$\mathbf{E} (\mathbf{E} (Z | \mathscr{F}_t) | \mathscr{F}_s) = \mathbf{E} (Z | \mathscr{F}_s) = X_s$$

and the claim follows. This is a very important class of martingales, actually by Theorem 5.3. every uniformly integrable martingale is of this type and we see from this Problem that we can find a lot of martingales given a filtration.

9. Let τ_1 and τ_2 be (\mathscr{F}_t) -stopping times. Show that

$$\tau_1 \wedge \tau_2, \quad \tau_1 \vee \tau_2, \quad \text{and} \quad \tau_1 + \tau_2$$

are (\mathscr{F}_t) -stopping times. (Hint: try to express the conditions as unions and intersections of conditions involving only τ_1 and τ_2 . Also discrete time versions are fine.) Suggestion.

We will show first that $\{\tau_1 \land \tau_2 \leq t\} \in \mathscr{F}_t$. If $\tau_1 \land \tau_2 \leq t$, then either $\tau_1 \leq t$ or $\tau_2 \leq t$ (or both). Moreover, if $\tau_1 \leq t$ or $\tau_2 \leq t$, then the minimum is also $\leq t$, so

$$\{\tau_1 \land \tau_2 \le t\} = \{\tau_1 \le t\} \cup \{\tau_2 \le t\} \in \mathscr{F}_t$$

since τ_1 and τ_2 are stopping times and therefore both of the events on the right-hand side and thus also their union is in \mathscr{F}_t .

We will show first that $\{\tau_1 \lor \tau_2 \leq t\} \in \mathscr{F}_t$. If $\tau_1 \lor \tau_2 \leq t$, then both $\tau_1 \leq t$ and $\tau_2 \leq t$. Moreover, if both $\tau_1 \leq t$ and $\tau_2 \leq t$, then the maximum is also $\leq t$, so

$$\{\tau_1 \lor \tau_2 \le t\} = \{\tau_1 \le t\} \cap \{\tau_2 \le t\} \in \mathscr{F}_t$$

since τ_1 and τ_2 are stopping times and therefore both of the events on the right-hand side and thus also their union is in \mathscr{F}_t .

Let's do the discrete time version first and then the general case.

Now $\tau_1 \in \{t_1, ..., t_n\}$ and $\tau_2 \in \{s_1, ..., s_m\}$ for some $0 \le t_1 < \cdots < t_n$ and $0 \le s_1 < \cdots < s_m$.

Now $\tau_1 + \tau_2 \in \{ t_j + s_k : j, k \}$ and $\tau_1 + \tau_2 \leq t$, when ever $\tau_1 = t_j$ and $\tau_2 = s_k$ satisfy the extra conditions $0 \leq t_j \leq t$ following $0 \leq s_k \leq t - t_j$. So

$$\{\tau_1 + \tau_2 \le t\} = \bigcup_{t_j \le t} \bigcup_{s_k \le t - t_j} \{\tau_1 = t_j, \tau_2 = s_k\}$$

For every fixed $t_j \leq t$ and $s_k \leq t - t_j$ the event on the right is in $\mathscr{F}_u \subset \mathscr{F}_t$, where $u = t_j \lor s_k \leq t$. So everything on the right is in \mathscr{F}_t so the claim follows in the discrete case for simple stopping times.

We need to reformulate the right-hand side for the general case since we cannot expect to have equalities, since only inequalities generalise. First we take the union over s_k for fixed t_j and we notice that

$$\{\tau_1 + \tau_2 \le t\} = \bigcup_{t_j \le t} \{\tau_1 = t_j, \tau_2 \le t - t_j\}$$

So now τ_2 is fine. Next we notice that if $\tau_2 \leq t - t_j$, then any $\tau_1 = t_l$ for $t_l \leq t_j$ must actually appear on the right, since $\tau_2 + \tau_1 \leq t - t_j + t_l < t$. So we can write the above as

$$\{\tau_1 + \tau_2 \le t\} = \bigcup_{t_j \le t} \{\tau_1 \le t_j, \tau_2 \le t - t_j\}$$

This is a good formulation for generalization and first we good try

$$\{\tau_1 + \tau_2 \le t\} = \bigcup_{s \le t} \{\tau_1 \le s, \tau_2 \le t - s\}$$

We notice that both sides are equal and for every fixed $s \leq t$ the event on the righthand side is in \mathscr{F}_t , but there is one issue left and that's the uncountability of the union.

But, fortunately, this is almost easy to fix, namely by restricting $s \in \mathbb{Q}$. Then

$$\{\tau_1 + \tau_2 \le t\} = \bigcup_{s < t, s \in \mathbb{Q}} \{\tau_1 \le s, \tau_2 \le t - s\} \cup \{\tau_1 = t, \tau_2 = 0\}$$

holds as well, since now the right-hand side is by previous a subset of $\{\tau_1 + \tau_2 \leq t\}$, but if $\tau_1(\omega) + \tau_2(\omega) \leq t$, then either $\tau_1(\omega) = t$ or $\tau_1(\omega) < t$. The former case is the special case on the right and in the latter case we can find a rational s such that $\tau_1(\omega) \leq s < t$ and $\tau_2(\omega) + s < t$. Why? Well the first condition says that $s \in [\tau_1(\omega), t)$ and the latter says that $s \in (-\infty, t - \tau_2(\omega))$. These intervals intersect if and only if $t - \tau_2(\omega) > \tau_1(\omega)$ which is equivalent with $\tau_1(\omega) + \tau_2(\omega) < t$.

Therefore, we have

$$\{\tau_1 + \tau_2 \le t\} = \bigcup_{s < t, s \in \mathbb{Q}} \{\tau_1 \le s, \tau_2 \le t - s\} \cup \{\tau_1 = t, \tau_2 = 0\}$$

and the right-hand side is a countable union of events in \mathscr{F}_t .

10. Show that the \mathscr{F}_{τ} is a σ -algebra, when τ is a (\mathscr{F}_t) -stopping time. **Suggestion**. Now by definition $\mathscr{F}_{\tau} = \{ A : \forall t \geq 0 : A \cap \{\tau \leq t\} \in \mathscr{F}_t \}$. We prove this by verifying that \mathscr{F}_{τ} satisfies the axioms of the σ -algebra.

- i) $\Omega \in \mathscr{F}_{\tau}$, since $\Omega \cap \{\tau \leq t\} = \{\tau \leq t\}$ and this is in \mathscr{F}_t since τ is a proper stopping time.
- *ii*) Suppose $A \in \mathscr{F}_{\tau}$. Then

$$\{A^{C}, \tau \le t\} = \{\tau \le t\} \setminus \{A, \tau \le t\} = \{\Omega, \tau \le t\} \cap \{A, \tau \le t\}^{C}.$$

Now the events on the right-hand side are in \mathscr{F}_t since $\Omega, A \in \mathscr{F}_\tau$ so we deduce that $A^C \in \mathscr{F}_\tau$.

iii) Suppose $(A_k) \subset \mathscr{F}_{\tau}$. Since

$$\{\tau \le t\} \cap \bigcup_k A_k = \bigcup_k \{A_k, \tau \le t\}$$

and by assumption everything on the right is in \mathscr{F}_t we have that $\bigcup A_k \in \mathscr{F}_{\tau}$ as well.

11. Let (\mathscr{F}_t) be a filtration. Show that the (\mathscr{F}_{t^+}) is right-continuous filtration. Suggestion. To show that $(\mathscr{G}_t) = (\mathscr{F}_{t^+})$ is right-continuous filtration, we need to show that

$$\bigcap_{s>t}\mathscr{G}_s=\mathscr{G}_t$$

Clearly, $\mathscr{G}_t \subset \bigcap \mathscr{G}_s$ if we know that (\mathscr{G}_t) is a filtration. So let's suppose $A \in \bigcap \mathscr{G}_s$. By definition, $A \in \mathscr{F}_{s^+}$ for every s > t. Which by definition means that $A \in \mathscr{F}_u$ for every u > s and every s > t. But this implies that $A \in \mathscr{F}_u$ for every u > t and therefore, $A \in \mathscr{F}_{t^+} = \mathscr{G}_t$. So, at least if we know that (\mathscr{G}_t) is a filtration, then it is right-continuous.

For completeness, let's show that \mathscr{G}_t is a σ -algebra.

- i) The $\Omega \in \mathscr{G}_t$ is quite easy to verify, since $\Omega \in \mathscr{F}_s$ for every s, and therefore, also in every \mathscr{F}_s for s > t.
- *ii*) Next, if $A \in \mathscr{G}_t$, then $A \in \mathscr{F}_s$ for every s > t. Therefore, $A^C \in \mathscr{F}_s$ for every s > t which means that $A^C \in \mathscr{G}_t$.
- *iii*) Next, if $(A_k) \subset \mathscr{G}_t$, then $(A_k) \subset \mathscr{F}_s$ for every s > t. Therefore, $A = \bigcup A_k \in \mathscr{F}_s$ for every s > t which means that $A \in \mathscr{G}_t$.

Let's still verify that $\mathscr{G}_t \subset \mathscr{G}_u$ for every t < u. If $A \in \mathscr{G}_t$, then as above $A \in \mathscr{F}_s$ for every s > t. Especially, $A \in \mathscr{F}_u \subset \mathscr{G}_u$.

12. Show that a random variable τ is a (\mathscr{F}_{t^+}) -stopping time if and only if for every t > 0 it holds that $\{\tau < t\} \in \mathscr{F}_t$. (Hint. \Longrightarrow consider events $\{\tau \le t - 1/k\}$ and \Leftarrow consider events $\{\tau < t + 1/k\}$.) Suggestion.

 \implies If τ is (\mathscr{F}_{t^+}) -stopping time, then $\{\tau \leq s\} \in \mathscr{F}_{s^+}$ for every s. Following the hint, let's take $s = t - \frac{1}{k}$. Then

$$\{\tau \le t - \frac{1}{k}\} \in \mathscr{F}_{s^+} \subset \mathscr{F}_t$$

for every k > 0. The second part $\mathscr{F}_{s^+} \subset \mathscr{F}_t$ follows, since $t > t - \frac{1}{k} = s$ and hence every $A \in \mathscr{F}_{s^+}$ must belong to \mathscr{F}_t as well. Moreover,

$$\{\tau < t\} = \bigcup_k \{\tau \le t - \frac{1}{k}\}$$

so we have deduced that $\{\tau < t\} \in \mathscr{F}_t$ as claimed.

 \Leftarrow Suppose now that $\{\tau < s\} \in \mathscr{F}_s$ for every s > 0. Following the hint, take $s_k = t + \frac{1}{k}$. This implies that

$$B_k := \{\tau < t + \frac{1}{k}\} \in \mathscr{F}_{s_k}.$$

Now

$$B := \{ \tau \le t \} = \bigcap_{k=1}^{\infty} \{ \tau < t + \frac{1}{k} \} = \bigcap_{k=1}^{\infty} B_k,$$

so we need to show that $B \in \mathscr{F}_{t^+}$. Since $\{\tau < t + \frac{1}{k+1}\} \subset \{\tau < t + \frac{1}{k}\}$, we see that (B_k) is *decreasing* sequence of events. This implies that

$$B = \bigcap_{k=1}^{\infty} B_k = \bigcap_{k=M}^{\infty} B_k$$

for every $M \ge 1$, since the clearly the left-hand side is a subset of the righthand side, but if $\omega \in B_k$ for every $k \ge M$, then it also belongs to $\omega \in B_M \subset B_{M-1} \subset \cdots \subset B_1$.

So let's assume that $k \geq M$. Then $s_k = t + \frac{1}{k} \leq t + \frac{1}{M} = s_M$, and so $B_k \in \mathscr{F}_{s_k} \subset \mathscr{F}_{s_M}$. This implies that

$$B = \bigcap_{k=M}^{\infty} B_k \in \mathscr{F}_{s_M}$$

for every $M \ge 1$. If s > t, then we can find $M \ge 1$ so that $s > s_M > t$ and so, $B \in \mathscr{F}_s$ as well. But this means that

 $B\in\mathscr{F}_s$

for every s > t or in other words $B = \{\tau \leq t\} \in \mathscr{F}_{t^+}$. This proves the claim.