Instructions. You are allowed to work on solutions in groups, but you are required to write up solutions on your own. Please give complete solutions, all claims need to be justified. Late homework will not be accepted. Please let me know if you find any misprints or mistakes.

1. Due by Feb 27, 11:00am

1. Let $(X_n)_{n \in \mathbb{N}}$ be an i.i.d. positive sequence, and $S_n = X_1 + \ldots + X_n$. Let $N_t = \sup\{n : S_n \leq t\}$. Prove that $S_{N_t} = S_{\lfloor t \rfloor}$ and $(N_t)_{t \in \mathbb{R}_+}$ are stochastic processes.

2. Let $v_1, v_2 : \mathbb{R}^2 \to \mathbb{R}^2$ be $C^\infty$ vector fields on $\mathbb{R}^2$. Let $N_t$ be defined as in the previous problem. Let $i(n) = 1$ if $n$ is odd and $i(n) = 2$ if $n$ is even.

Let $Z_t$ denote the solution of the ODE $\frac{d}{dt}Z_t = v_{i(N_t)}(Z_t), \quad Z_0 = x$.

Prove that $(Z_t)_{t \geq 0}$ is a stochastic process.

3. Prove that cylinders $C(t_1, \ldots, t_n, B), B \in \mathcal{B}(\mathbb{R}^n), t_1, \ldots, t_n \geq 0, n \in \mathbb{N}$ form an algebra.

4. We can define the cylindrical $\sigma$-algebra as the $\sigma$-algebra generated by elementary cylinders or by cylinders. Prove that these definitions are equivalent.

5. Let $\mathcal{F}_T = \sigma\{C(t_1, \ldots, t_n, B) : t_1, \ldots, t_n \in T\}$ for $T \subset \mathbb{T}$.

Prove that

$$\mathcal{B}(\mathbb{R}^T) = \bigcup_{\text{countable } T \subset \mathbb{T}} \mathcal{F}_T.$$  

6. Prove that $X : \mathbb{T} \times \Omega \to \mathbb{R}$ is a stochastic process iff $X$ seen as $X : \Omega \to \mathbb{R}^T$ is $(\Omega, \mathcal{F}) \to (\mathbb{R}^T, \mathcal{B}(\mathbb{R}^T))$ measurable.

7. Use characteristic functions to prove the existence of a Wiener process (up to continuity of paths).

8. Let $(X_t)_{t \in [0,1]}$ be an (uncountable) family of i.i.d. r.v.’s with nondegenerate distribution. Prove that no modification of this process can be continuous.

9. A multidimensional version of the Kolmogorov–Chentsov theorem. Suppose $d \geq 1$, and there is a stochastic field $X : [0,1]^d \times \Omega \to \mathbb{R}$ that satisfies $E|X(s) - X(t)|^\alpha \leq C|s - t|^{d+\beta}$ for some $\alpha, \beta, C > 0$ and all $t, s \in [0,1]^d$.

Prove that there is a continuous modification of $X$ on $[0,1]^d$. 
10. Show that the Kolmogorov–Chentsov theorem cannot be relaxed: inequality $E|X_t - X_s| \leq C|t - s|$ is not sufficient for existence of a continuous modification. Hint: consider the following process: let $\tau$ be a r.v. with exponential distribution, and define $X_t = 1_{\{\tau \leq t\}}$.

11. Prove that there exists a Poisson process (a process with independent increments that have Poisson distribution with parameter proportional to the length of time increments) such that:
   (a) its realizations are nondecreasing, taking only whole values a.s.
   (b) its realizations are continuous on the right a.s.
   (c) all the jumps of the realizations are equal to 1 a.s.

12. Give an example of a non-Gaussian 2-dimensional random vector with Gaussian marginal distributions.

13. Let $Y \sim \mathcal{N}(a, C)$ be a $d$-dimensional random vector. Let $Z = AY$ where $A$ is an $n \times d$ matrix. Prove that $Z$ is Gaussian and find its mean and covariance matrix.

14. Prove that for every vector $b \in \mathbb{R}^d$, the r.v. $\langle b, X \rangle$ is Gaussian.

15. Prove that $\langle 1_{[0,t]}, 1_{[0,s]} \rangle_{L^2(\mathbb{R}^+)} = t \wedge s$.

16. Use the Chentsov–Kolmogorov theorem to find a condition on the mean $a(t)$ and covariance function $c(s,t)$ that guarantees existence of a continuous Gaussian process with these parameters.

17. Suppose $(X_0, X_1, \ldots, X_n)$ is a (not necessarily centered) Gaussian vector. Show that there are constants $c_0, c_1, \ldots, c_n$ such that
   $$\mathbb{E}(X_0|X_1, \ldots, X_n) = c_0 + c_1 X_1 + \ldots + c_n X_n.$$
function \( f : \mathbb{R}_+ \to \mathbb{R}_+ \) such that \( X \) has the same f.d.d.'s as \( Y \) defined by 
\[
Y(t) = W(f(t)),
\]
for a Wiener process \( W \).

2. DUE BY MARCH 27, 11:00AM

1. Let \( \mu \) be a \( \sigma \)-finite Borel measure on \( \mathbb{R}^d \). Prove existence of Poisson point process and (Gaussian) white noise with leading measure \( \mu \).

   In both cases, the process \( X \) we are interested in is indexed by Borel sets \( A \) with \( \mu(A) < \infty \), with independent values on disjoint sets, with finite additivity property for disjoint sets and such that
   - \( X(B) \) is Poisson with parameter \( \mu(B) \) (for Poisson point process).
   - \( X(B) \) is centered Gaussian with variance \( \mu(B) \) (for white noise).

2. Suppose the process \( X_t \) is a Gaussian process, and let \( H \) be the Hilbert space generated by \((X_t)_{t \in \mathbb{R}}\), i.e., the space consisting of \( L^2 \)-limits of linear combinations of values of \( X_t \). Prove that every element in \( H \) is a Gaussian r.v.

3. Let \( A, \eta \) be random variables, let a r.v. \( \phi \) be independent of \((A, \eta)\) and uniform on \([0, 2\pi]\). Prove that \( X_t = A \cos(\eta t + \phi) \), \( t \in \mathbb{R} \), is a strictly stationary process.

4. Find the covariance function of a stationary process such that its spectral measure is \( \rho(dx) = \frac{dx}{1+x^2} \).

5. Give an example of a weakly stationary stochastic process \((X_n)_{n \in \mathbb{N}}\) such that \((X_1 + \ldots + X_n)/n\) converges in \( L^2 \) to a limit that is not a constant. Explain what is going on in terms of the spectral representation.

6. Let \((X_n)_{n \in \mathbb{Z}}\) be a weakly stationary process. Prove that for any \( K \in \mathbb{N} \) and any numbers \( a_{-K}, a_{-K+1}, \ldots, a_{K-1}, a_K \), the process \((Y_n)_{n \in \mathbb{Z}}\) defined by
   \[
   Y_n = \sum_{k=-K}^{K} a_k X_{n+k}
   \]
is weakly stationary. Express the spectral measure of \( Y \) in terms of the spectral measure for \( X \).

7. Describe the stationary sequence
   \[
   X_n = \int_0^{2\pi} \cos(n\lambda)Z(d\lambda), \quad n \in \mathbb{Z},
   \]
where \( Z(\cdot) \) is the standard white noise on \([0, 2\pi]\).

8. Let a stationary process \((X_n)_{n \in \mathbb{Z}}\) satisfy \( \mathbb{E}|X_0| < \infty \). Prove that with probability 1, \( \lim_{n \to \infty} (X_n/n) = 0 \).

9. Find the spectral measure and spectral representation for stationary process \((X_t)_{t \in \mathbb{R}}\) given by
   \[
   X_t = A \cos(t + \phi), \quad t \in \mathbb{R},
   \]
where \( A \) and \( \phi \) are independent, \( A \in L^2 \) and \( \phi \) is uniform on \([0, 2\pi]\).

10. Consider a map \( \theta : \Omega \to \Omega \). A set \( A \) is called (backward) invariant if \( \theta^{-1}A = A \), forward invariant if \( \theta A = A \). Prove that the collection of
backward invariant sets forms a \( \sigma \)-algebra. Give an example of \( \Omega \) and \( \theta \) such that the collection of forward invariant sets does not form a \( \sigma \)-algebra.

11. Consider the transformation \( \theta : \omega \mapsto \{ \omega + \lambda \} \) on \( \Omega = [0, 1) \) equipped with Lebesgue measure. Here \( \{ \ldots \} \) denotes the fractional part of a number. This map can be interpreted as rotation of the circle (seen as \([0, 1)\) with endpoints 0 and 1 identified).

A. Prove that this dynamical system is ergodic (i.e., there are no backward invariant Borel sets besides \( \emptyset \) and \( \Omega \)) if and only if \( \lambda \not\in \mathbb{Q} \). Hint: for \( \lambda \in \mathbb{Q} \) construct a backward invariant subset, for \( \lambda \not\in \mathbb{Q} \) take the indicator of an invariant set and write down the Fourier series for it (w.r.t. \( e^{2\pi i n \omega} \)). What happens to this expansion under \( \theta \)?

B. Using the Birkhoff ergodic theorem and Part A, describe the limiting behavior of \( (f(\omega) + f(\theta^1 \omega) + \ldots + f(\theta^{n-1} \omega))/n \) as \( n \to \infty \), where \( f \) is any \( L^1 \) function on \([0,1]\).

12. Prove that every Gaussian martingale is a process with independent increments.

13. Prove that if \( W_t \) is a standard Brownian motion, then \( e^{\sigma W_t - \frac{\sigma^2 t}{2}} \) is a martingale for any \( \sigma \in \mathbb{R} \).

14. Prove that if \( W_t \) is a standard Brownian motion, then \( W_t^3 - 3 \int_0^t W_s ds \) is a martingale.

15. Show that the function

\[
P(s, x, t, \Gamma) = P(t-s, x, \Gamma) = \frac{1}{\sqrt{2\pi(t-s)}} e^{-\frac{(x-y)^2}{2(t-s)}} dy
\]

is a Markov transition probability function for the standard Wiener process.

16. Let \( X_t = t^\alpha + W_t \), where \( (W_t) \) is the standard Wiener process. Find necessary and sufficient conditions on \( \alpha \in \mathbb{R} \) to ensure that \( P\{\lim_{t \to \infty} X_t = +\infty\} = 1 \).

17. Let \( (X_t)_{t \geq 0} \) be a stochastic process. We define \( \mathcal{F}_{\leq t} = \sigma(X_s, 0 \leq s \leq t) \) and \( \mathcal{F}_{\geq t} = \sigma(X_s, s \geq t) \).

Prove that the following two versions of the Markov property are, in fact, equivalent:

1. For all \( A \in \mathcal{F}_{\leq t}, B \in \mathcal{F}_{\geq t}, P(AB|X_t) = P(A|X_t)P(B|X_t) \).
2. For all \( B \in \mathcal{F}_{\geq t}, P(B|\mathcal{F}_{\leq t}) = P(B|X_t) \).

18. Let \( W \) be a Wiener process. The previous problem shows that the process \( X_t = W_{-t}, t \in (-\infty, 0] \) is Markov. Find the transition probability function for \( X_t \).

19. Prove that the process \( (X_t) \) given by \( X_t = |W_t| \) is Markov. Find its transition probability function.

20. Let \( (N_t)_{t \geq 0} \) be the standard Poisson process. Let \( X_t = (-1)^{N_t} \). Prove that \( (X_t) \) is a Markov process and find the transition probability function.

21. We say that a r.v. \( \tau \) is a stopping time w.r.t. a filtration \( (\mathcal{F}_t)_{t \geq 0} \) if for every \( t, \{\tau \leq t\} \in \mathcal{F}_t \).
Let \((X_t, \mathcal{F}_t)_{t \geq 0}\) be a continuous process such that \(X_0 = 0\). Suppose \(a > 0\) and let \(\tau = \inf\{t : X(t) > a\}\) and \(\nu = \inf\{t : X(t) \geq a\}\). Show that \(\nu\) is a stopping time w.r.t. \((\mathcal{F}_t)_{t \geq 0}\) and \(\tau\) is a stopping time w.r.t. \((\mathcal{F}_{t+})_{t \geq 0}\), where \(\mathcal{F}_{t+} = \bigcap_{\varepsilon > 0} \mathcal{F}_{t+\varepsilon}\).

22. Show that if \(\tau_1 \leq \tau_2 \leq \ldots\) are stopping times w.r.t. to a filtration \((\mathcal{F}_t)\), then \(\tau = \lim_{n \rightarrow \infty} \tau_n\) is also a stopping time w.r.t. to \((\mathcal{F}_t)\).

23. Let \(\mathcal{F}_\tau = \{A : A \cap \{\tau \leq t\} \in \mathcal{F}_t\}\) for a filtration \((\mathcal{F}_t)\) and a stopping time \(\tau\). Show that \(\mathcal{F}_\tau\) is a \(\sigma\)-algebra.

24. Give an example of the following: a random variable \(\tau \geq 0\) is not a stopping time, \(\mathcal{F}_\tau\) is not a \(\sigma\)-algebra.

25. Suppose \(\tau\) is a stopping time w.r.t. \((\mathcal{F}_{t+})\). Let us define
\[
\tau_n = \frac{[2^n \tau] + 1}{2^n} = \sum_{k \in \mathbb{N}} \frac{k}{2^n} \mathbb{1}_{\{\tau \in [\frac{k}{2^n}, \frac{k+1}{2^n})\}}, \quad n \in \mathbb{N}.
\]
Prove that for every \(n \in \mathbb{N}\), \(\tau_n\) is a stopping time w.r.t. \((\mathcal{F}_t)_{t \geq 0}\), \(\mathcal{F}_{\tau_n} \supset \mathcal{F}_{\tau+}\), and \(\tau_n \downarrow \tau\).

26. Prove: if \((X_t, \mathcal{F}_t)_{t \geq 0}\) is a continuous process, then for any stopping time \(\tau\), \(X_\tau\) is a r.v. measurable w.r.t. \(\mathcal{F}_\tau\).

3. DUE BY APRIL 24, 11:00AM

1. For a Wiener process \(W\) and \(b > 0\), compute the density of
\[
\tau_b = \inf\{t \geq 0 : W_t = b\}
\]
and find \(\mathbb{E}\tau_b\).

2. Let \((X_t, \mathcal{F}_t)\) be a continuous martingale and let \(\tau\) be a stopping time w.r.t. \(\mathcal{F}_\tau\). Prove that the “stopped” process \((X^\tau_t, \mathcal{F}_t)_{t \geq 0}\), where \(X^\tau_t = X_{\tau \wedge t}\), is also a martingale.

3. Prove the following theorem (Kolmogorov, 1931) using Taylor expansions of test functions:

Suppose \((P_x)_{x \in \mathbb{R}^d}\) is a (homogeneous) Markov family on \(\mathbb{R}^d\) with transition probabilities \(P(\cdot, \cdot, \cdot)\). Suppose there are continuous functions \(a^{ij}(x)\), \(b^i(x)\), \(i, j = 1, \ldots, d\), such that for every \(\varepsilon > 0\), the following relations hold uniformly in \(x\):
\[
P(t, x, B^c_\varepsilon(x)) = o(t), \quad t \rightarrow 0,
\]
\[
\int_{B_\varepsilon(x)} (y^i - x^i) P(t, x, dy) = b^i(x)t + o(t), \quad t \rightarrow 0,
\]
\[
\int_{B_\varepsilon(x)} (y^i - x^i)(y^j - x^j) P(t, x, dy) = a^{ij}(x)t + o(t), \quad t \rightarrow 0.
\]
where \(B_\varepsilon(x)\) is the Euclidean ball of radius \(\varepsilon\) centered at \(x\).

Then the infinitesimal generator \(A\) of the Markov semigroup associated to the Markov family is defined on all functions \(f\) such that \(f\) itself and all
its partial derivatives of first and second order are bounded and uniformly continuous. For such functions

$$Af = \frac{1}{2} \sum_{ij} a^{ij} \partial_{ij} f + \sum_i b^i \partial_i f.$$ 

4. Consider a Markov process $X$ in $\mathbb{R}^2$ given by

$$X_1(t) = X_1(0) + W(t),$$
$$X_2(t) = X_2(0) + \int_0^t X_1(s) ds.$$ 

Find its generator on $C^2$-functions with compact support.

5. Consider the Poisson transition probabilities, i.e., fix a number $\lambda > 0$ and for $i \in \mathbb{Z}$ and $t \geq 0$, let $P(i, t, \cdot)$ be the distribution of $i + \pi_t$, where $\pi_t$ denotes a random variable with Poisson distribution with parameter $s > 0$. In other words,

$$P(i, t, \{j\}) = e^{-\lambda t} \frac{(\lambda t)^j}{j!}, \quad i \in \mathbb{Z}, \quad j \in \{i, i+1, \ldots\}, \quad t > 0.$$ 

Find the generator of the Markov semigroup on all bounded test functions $f : \mathbb{Z} \to \mathbb{R}$.

6. Find the transition probabilities and generator associated to the Ornstein-Uhlenbeck process.

7. Let $W^1$ and $W^2$ be two independent Wiener processes w.r.t. a filtration $(\mathcal{F}_t)_{t \geq 0}$, and let $X$ be a bounded process adapted to $(\mathcal{F}_t)_{t \geq 0}$.

For a partition $t$ of time interval $[0, T]$ (i.e., a sequence of times $t = (t_0, t_1, \ldots, t_n)$ such that $0 = t_0 < t_1 < t_2 < \ldots < t_n = T$), we define

$$Q(t) = \sum_j X_{t_j} (W^1_{t_{j+1}} - W^1_{t_j}) (W^2_{t_{j+1}} - W^2_{t_j}).$$

Prove:

$$\lim_{\max(t_{j+1} - t_j) \to 0} Q(t) = 0 \quad \text{in } L^2.$$ 

(In particular, $(W^1, W^2)_{t \geq 0}$ is a Wiener process.

8. Let $(\mathcal{F}_t)$ be a filtration. Suppose that $0 = A_0(t) + A_1(t) W(t)$ for all $t$, where $(A_0, \mathcal{F}_t)$ and $(A_1, \mathcal{F}_t)$ are processes with $C^1$ trajectories, and $W(t)$ is a Wiener process w.r.t. $(\mathcal{F}_t)$. Prove that $A_0 \equiv 0$ and $A_1 \equiv 0$.

9. Suppose $f \in L^2([0, T], \mathcal{B}, \text{Leb})$. For all $h > 0$, we define

$$f_h(t) = \begin{cases} 
0, & 0 \leq t < h; \\
\frac{1}{h} \int_{(k-1)h}^{kh} f(s) ds, & kh \leq t < (k+1)h \wedge T.
\end{cases}$$

Prove that $f_h \in L^2$ and $\|f_h\|_{L^2} \leq \|f\|$. Hint: do not work with generic $L^2$ functions right away; consider a smaller supply of functions $f$ that behave nicely under this kind of averaging.
10. Compute $\int_0^T W^2(t) dW(t)$ directly as the limit of
\[ \sum W^2(t_k)(W(t_{k+1}) - W(t_k)) \]

11. The so-called Stratonovich stochastic integral may be defined for a broad class of adapted processes $X_t$ via
\[ \int_0^T X_t \circ dW_t \overset{def}{=} \lim_{\max(t_{j+1}-t_j)\to 0} \sum X_{t_j+1} + X_{t_j} \frac{W_{t_{j+1}} - W_{t_j}}{2} \text{ in } L^2. \]

Impose any conditions you need on $X$ and express the difference between the Itô and Stratonovich integrals in terms of quadratic covariation between $X$ and $W$. Compute $\int_0^T W_t \circ dW_t$. Is the answer a martingale?

12. Let $\mathcal{M}_2^c = \{\text{square-integrable martingales with continuous paths}\}$. Prove that if $(M_t, F_t) \in \mathcal{M}_2^c$, then
\[ \mathbb{E}[(M_t - M_s)^2 | F_s] = \mathbb{E}[M_t^2 - M_s^2 | F_s] = \mathbb{E}[\langle M \rangle_t - \langle M \rangle_s | F_s], \quad s < t. \]

13. Suppose $(M_t, F_t) \in \mathcal{M}_2^c$, $X$ is a simple process, and $(X \cdot M)_t = \int_0^t X_s dM_s$. Prove that
\[ \mathbb{E}[(X \cdot M)_t - (X \cdot M)_s)^2 | F_s] = \mathbb{E} \left[ \int_s^t X_r^2 d\langle M \rangle_r | F_s \right], \quad s < t. \]

14. Let $M \in \mathcal{M}_2^c$. Prove that
\[ Y \cdot (X \cdot M) = (YX) \cdot M \]
for simple processes $X, Y$. Find reasonable weaker conditions on $X$ and $Y$ guaranteeing the correctness of this identity in the sense of square integrable martingales.

15. Suppose $(M_t, F_t) \in \mathcal{M}_2^c$, and $X, Y$ are bounded processes. Prove that
\[ \langle X \cdot M, Y \cdot M \rangle_t = \int_0^t X_s Y_s d\langle M \rangle_s. \]

Here, for two processes $M, N \in \mathcal{M}_2^c$ the cross-variation $\langle M, N \rangle_t$ is defined by
\[ \langle M, N \rangle_t = \frac{(M + N)_t - (M - N)_t}{4}. \]

16. Let us define the process $X$ by
\[ X_t = e^{\lambda t} X_0 + \varepsilon e^{\lambda t} \int_0^t e^{-\lambda s} dW_s, \quad t \geq 0. \]

Here $\lambda \in \mathbb{R}$, $\varepsilon > 0$, $W$ is a standard Wiener process, and $X_0$ is a square-integrable r.v., independent of $W$. Prove that
\[ dX_t = \lambda X_t dt + \varepsilon dW_t. \]
17. Prove that if \( f : [0, \infty) \) is a deterministic function, bounded on any interval \([0, t]\), then
\[
X_t = \int_0^t f(s) dW_s, \quad t \geq 0,
\]
is a Gaussian process. Find its mean and covariance function.

18. In the context of Problem 16, find all the values of \( \lambda \) with the following property: there are \( a \) and \( \sigma^2 \) such that if \( X_0 \sim N(a, \sigma^2) \), then \( (X_t) \) is a stationary process.

19. [Feynman–Kac formula.] Suppose \( u_0 : \mathbb{R} \to [0, \infty) \) and \( \phi : [0, \infty) \times \mathbb{R} \to \mathbb{R} \) are smooth bounded functions. Suppose that \( u : [0, \infty) \times \mathbb{R} \to \mathbb{R} \) is a smooth function solving the Cauchy problem for the following equation:
\[
\partial_t u(t, x) = \frac{1}{2} \partial_{xx} u(t, x) + \phi(t, x) u(t, x),
\]
\[
u(0, x) = u_0(x).
\]
Using the Itô formula and martingales, prove that
\[
u(t, x) = \mathbb{E} \left[ f(x + W_t) + \int_0^t \phi(s, x + W_s) ds \right], \quad t > 0, \ x \in \mathbb{R},
\]
where \( W \) is a standard Wiener process. (If you need further assumptions on \( u \) and/or its derivatives, use PDE theory or just impose assumptions you need.)

4. DUE BY MAY 15, 11:00 AM.

Please send me solutions by email. Acceptable formats are PDF and popular image formats like JPEG (type in \LaTeX{} and compile or scan / take pictures of your handwritten notes).

1. Let \( D \) be an open bounded of \( \mathbb{R}^d \) with smooth boundary \( \partial D \). Let \( g : D \to \mathbb{R} \) and \( f : \partial D \to \mathbb{R} \) be continuous functions. Suppose that \( u \in C(\overline{D}) \cap C^2(D) \) and
\[
\frac{1}{2} \Delta u(x) = -g(x), \quad x \in D,
\]
\[
u(x) = f(x), \quad x \in \partial D,
\]
where \( \Delta f(x_1, x_2) = \partial^2_{11} f(x_1, x_2) + \partial^2_{22} f(x_1, x_2) \) is the Laplace operator.
Let \( W \) be a standard 2-dimensional Wiener process (a process with independent components each of which is a standard 1-dimensional Wiener process)

Applying the Ito formula to the process \( X_t = u(x + W_t) + \int_0^t g(x + W_s) ds \), prove that if \( \tau = \inf \{ t \geq 0 : x + W_t \in \partial D \} \), then
\[
u(x) = \mathbb{E} \left[ f(x + W_\tau) + \int_0^\tau g(x + W_t) dt \right].
2. Let $W$ be the standard Brownian Motion. Prove that \( P\{\sup_{t<T}|W_t|<1\} = u(T,0) \), where $u = u(t,x)$ is the solution of the following PDE:

\[
\begin{align*}
\partial_t u(t,x) &= \frac{1}{2} \partial_{xx} u(t,x), \quad (t,x) \in (0,\infty) \times (-1,1) \\
u(0,x) &= 1, \\
u(t,\pm 1) &= 0.
\end{align*}
\]

Use this to find $C, \lambda > 0$ such that

\[P\{\sup_{t<T}|W_t|<1\} = Ce^{-\lambda T}(1+o(1))\]

as $T \to +\infty$.

3. Suppose $a \in \mathbb{R}$, $\sigma > 0$, $x_0 > 0$, and $W$ is the standard Wiener process. Find constants $A, B \in \mathbb{R}$ such that the process $S$ defined for all $t \geq 0$ by $S_t = x_0 \exp(at + \sigma W_t)$ (and often called "the geometric Brownian motion") satisfies the following stochastic equation

\[dS_t = AS_t dt + BS_t dW_t, \quad t \geq 0.\]

Find necessary and sufficient conditions on $a$ and $\sigma$ for $(S_t)$ to be a martingale.

4. [Exponential martingale inequality] Prove that for any continuous local martingale $M$,

\[P\left\{ \sup_{t \geq 0} |M_t| > x, \langle M \rangle_{\infty} < y \right\} \leq 2e^{-x^2/2y}.
\]

5. Consider the following family of processes indexed by $n \in \mathbb{N}$:

\[X_n(t) = \int_0^{n+t} e^{s-(n+t)} dW_s, \quad t \in [0,1].\]

Here $W$ is a standard Brownian motion. Prove that as $n \to \infty$, $X_n$ converges in distribution in $C[0,1]$. Describe the limiting distribution.

6. Suppose a sequence of continuous processes $X_n$ converges in distribution in $C[0,1]$ to the standard Brownian motion. Let

\[\tau_n = \inf\{t \geq 0 : |X_t| + t \geq 1\}.
\]

Is it true that $\tau_n$ converges in distribution to $\tau = \inf\{t \geq 0 : |W_t| + t \geq 1\}$?

7. Let $X_1, X_2, \ldots$ be an i.i.d. sequence of r.v.’s satisfying $EX_k = 0$, and $EX_k^2 = 1$. Let $S_n = X_1 + \ldots + X_n$ for all $n$. Let $S_n^* = \max\{0, S_1, S_2, \ldots, S_n\}$. Use Donsker’s Functional CLT to find the limiting distribution of $S_n^*/\sqrt{n}$.

8. Let $a > 0$ and let $X_t = W_t + at$ be the Brownian motion with drift $a$. The Girsanov theorem implies that the distribution of $X$ is equivalent to the Wiener measure. However, $P\{X_t \to +\infty\} = 1$ and $P\{W_t \to +\infty\} = 0$ (prove both identitites), so these measures are actually singular. Write the Girsanov density and explain what is going on.

9. Use Girsanov’s theorem to prove that if $b$ is a bounded measurable function, then for every $t > 0$, the distribution of $X(t)$, the solution of $X(t) = \int_0^t b(X(s)) dt + W(t)$, is equivalent to the Lebesgue measure.
10. The Bessel($d$) process

$$R_d(t) = \sqrt{(x_1 + W_1(t))^2 + \ldots + (x_d + W_d(t))^2}$$

with initial condition $r = \sqrt{x_1^2 + \ldots + x_d^2}$ is defined as the absolute value of $d$-dimensional Wiener process with standard independent components emitted from some point $x = (x_1, \ldots, x_d)$. Prove that there is a Wiener process $B$ such that

$$dR_d(t) = \frac{d - 1}{2R_d(t)} dt + dB(t).$$

Due to the previous problem, it may seem plausible that the distribution of $R_d(t)$ should be equivalent to the Lebesgue measure but $R_d$ is clearly nonnegative, so this equivalence does not hold. What is going on?