

Exercise 1.1, HW2 Stochastic, Mathew Rakha Pradyatama

$$1.1) f\left(\frac{j}{n}\right) = \frac{j^2}{n^2}, \quad f\left(\frac{j-1}{n}\right) = \frac{(j-1)^2}{n^2} = \frac{j^2 - 2j + 1}{n^2}$$

$$f\left(\frac{j}{n}\right) - f\left(\frac{j-1}{n}\right) = \frac{j^2 - (j^2 - 2j + 1)}{n^2} = \frac{2j - 1}{n^2}$$

$$Q = \lim_{n \rightarrow \infty} \sum_{j=1}^n \left| \frac{2j-1}{n^2} \right|^{5/4} = \lim_{n \rightarrow \infty} \sum_{j=1}^n \frac{(2j-1)^{5/4}}{n^{5/2}}$$

as $n \rightarrow \infty$,

$$Q \longrightarrow \frac{\sum_{j=1}^{\infty} (2j-1)^{5/4}}{\infty} = 0 //$$

as $n \rightarrow \infty$, Q goes to $0 //$

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Exercise 1.2

For a standard Brownian motion $B_t \sim N(0, t)$,

$$B_t - B_s \sim N(0, (t-s))$$

$$B_{\frac{j}{n}} - B_{\frac{j-1}{n}} \sim N\left(0, \frac{j}{n} - \frac{j-1}{n}\right) = N\left(0, \frac{1}{n}\right) \rightarrow \sigma^2 = \frac{1}{n}$$

$$\text{To scale } X \sim N(0, \sigma^2) \text{ to } Z \sim N(0, 1) \quad \left\{ \begin{array}{l} \sigma = \left(\frac{1}{n}\right)^{1/2} \end{array} \right.$$

$$Z = \frac{X - \mu}{\sigma} = \frac{X - 0}{\sigma} = \frac{X}{\sigma} \rightarrow X = \sigma \cdot Z$$

$$E[|X|^p] = E[|\sigma Z|^p] = \sigma^p \cdot E[|Z|^p]$$

$$\begin{aligned} E\left[|f\left(\frac{j}{n}\right) - f\left(\frac{j-1}{n}\right)|^{5/4}\right] &= E\left[|B\left(\frac{j}{n}\right) - B\left(\frac{j-1}{n}\right)|^{5/4}\right] = \sigma^{5/4} E[|Z|^{5/4}] \\ &= \left(\left(\frac{1}{n}\right)^{1/2}\right)^{5/4} \cdot E[|Z|^{5/4}] = \left(\frac{1}{n}\right)^{5/8} E[|Z|^{5/4}] \end{aligned}$$

where $Z \sim N(0, 1)$

$$E(Q) = \sum_{j=1}^n E\left[|f\left(\frac{j}{n}\right) - f\left(\frac{j-1}{n}\right)|^{5/4}\right] = (n) \left(\frac{1}{n}\right)^{5/8} E[|Z|^{5/4}]$$

$$E(Q) = n^{3/8} E[|Z|^{5/4}] \rightarrow \infty \text{ as } n \rightarrow \infty$$

$$Q = \lim_{n \rightarrow \infty} E(Q) = \infty //$$

Q diverges as $n \rightarrow \infty //$

Stochastic Calculus – Homework 2

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Exercise 2.1

$$Y_t = 2B_t - W_t - t$$

(Page 68 of class notes) There are 4 properties to prove:

1. $Y_0 = 0$
2. If $s < t$, the distribution of $Y_t - Y_s$ is joint normal with mean $m(t - s)$ and covariance matrix $(t - s)\Gamma$
3. If $s < t$, the random vector $Y_t - Y_s$ is independent of \mathcal{F}_s
4. With probability one, the function $t \rightarrow Y_t$ is continuous

Property 1

Since B_t is a standard Brownian motion, $B_0 = 0$

Since W_t is a standard Brownian motion, $W_0 = 0$

$$Y_0 = 2B_0 - W_0 - 0 = 2 * 0 - 0 = 0$$

We have proven that $Y_0 = 0$, meaning that Y_t starts at the origin the first property is satisfied.

Property 2

$$\begin{aligned} Y_t - Y_s &= 2B_t - W_t - t - 2B_s + W_s + s \\ Y_t - Y_s &= 2(B_t - B_s) - (W_t - W_s) - (t - s) \end{aligned}$$

The distribution of $(Y_t - Y_s)$ depends on the distribution of $(B_t - B_s)$ and $(W_t - W_s)$.

Since B_t is a standard Brownian motion, $B_t - B_s$ has variance $(t - s)$

Since W_t is a standard Brownian motion, $W_t - W_s$ has variance $(t - s)$

$(t - s)$ is a constant so it has a variance of 0

$$Var(Y_t - Y_s) = Var(2(B_t - B_s)) + Var(-(W_t - W_s)) + Var(-(t - s))$$

$$Var(Y_t - Y_s) = 4Var(B_t - B_s) + Var(W_t - W_s) + 0$$

$$Var(Y_t - Y_s) = 4(t - s) + (t - s) = 5(t - s)$$

$$E(Y_t - Y_s) = m(t - s) = -1(t - s)$$

Since Y_t is a one-dimensional Brownian motion, the covariance matrix is equal to the variance of Y_t .

Since the distributions of $(B_t - B_s)$ and $(W_t - W_s)$ are normal with mean 0 and variance $(t - s)$, the distribution of $(Y_t - Y_s)$ is joint normal with mean $-1(t - s)$ and covariance matrix $(t - s) * 5$. The second property is satisfied.

Property 3

$$Y_t - Y_s = 2(B_t - B_s) - (W_t - W_s) - (t - s)$$

The random vector $Y_t - Y_s$ depends on the values of $(B_t - B_s)$, $(W_t - W_s)$, and $(t - s)$.

Since B_t and W_t are Brownian motions, $(B_t - B_s)$ and $(W_t - W_s)$ don't depend on the information up to time s , i.e. \mathcal{F}_s .

$(B_t - B_s)$ and $(W_t - W_s)$ are independent of \mathcal{F}_s .

The value of $(t - s)$ is independent of \mathcal{F}_s .

Since all of the components are independent of \mathcal{F}_s , $Y_t - Y_s$ is also independent of \mathcal{F}_s . The third property is satisfied.

Property 4

Since B_t and W_t are standard Brownian motions, both are continuous functions of t . Since Y_t is merely a linear combination of B_t and W_t , a linear combination of continuous functions of t , is also a continuous function of t . Therefore, with probability one, the function $t \rightarrow Y_t$ is continuous. The fourth property is satisfied.

Conclusion: Since Y_t satisfies all 4 properties, Y_t is a one-dimensional Brownian motion starting at the origin ($Y_0 = 0$).

$$Y_t = 2B_t - W_t - t$$

Since B_t and W_t are standard Brownian motions, B_t and W_t has mean 0 and variance t is a constant so it has a variance of 0.

$$\begin{aligned} E(Y_t) &= E(2B_t) - E(W_t) - E(t) = 0 + 0 - t \\ E(Y_t) &= -t \\ m &= -1 \end{aligned}$$

$$\begin{aligned} Var(Y_t) &= Var(2B_t) + Var(-W_t) + Var(-t) = 4Var(B_t) + Var(W_t) + 0 = 4t + t \\ Var(Y_t) &= 5t \\ \sigma^2 &= 5 \end{aligned}$$

For Y_t , drift (m) is -1, variance parameter (σ^2) is 5.

Exercise 2.2

$$\begin{aligned}Z_t &= (Z_t^1, Z_t^2) \\Z_t^1 &= Y_t + t = 2B_t - W_t - t + t \\Z_t^1 &= 2B_t - W_t \\Z_t^2 &= -B_t + 3W_t\end{aligned}$$

Property 1

$$\begin{aligned}Z_0^1 &= 2B_0 - W_0 = 2 * 0 + 0 = 0 \\Z_0^2 &= -B_0 + 3W_0 = 0 + 3 * 0 = 0 \\Z_0 &= (Z_0^1, Z_0^2) = (0, 0)\end{aligned}$$

We have proven that $Z_0 = (0, 0)$, meaning that Z_t starts at the origin, and that the first property is satisfied.

Property 2

$$\begin{aligned}Z_t - Z_s &= (Z_t^1 - Z_s^1, Z_t^2 - Z_s^2) \\Z_t^1 - Z_s^1 &= 2B_t - W_t - (2B_s - W_s) \\Z_t^1 - Z_s^1 &= 2(B_t - B_s) - 1(W_t - W_s)\end{aligned}$$

The distribution of $(Z_t^1 - Z_s^1)$ depends on the distributions of $(B_t - B_s)$ and $(W_t - W_s)$.
The distributions of $(B_t - B_s)$ and $(W_t - W_s)$ are normal with mean 0 and variance $(t - s)$.

$$\begin{aligned}\text{Var}(Z_t^1 - Z_s^1) &= \text{Var}\{2(B_t - B_s)\} + \text{Var}\{-1(W_t - W_s)\} \\ \text{Var}(Z_t^1 - Z_s^1) &= 4\text{Var}(B_t - B_s) + 1\text{Var}(W_t - W_s) \\ \text{Var}(Z_t^1 - Z_s^1) &= 4(t - s) + 1(t - s) \\ \text{Var}(Z_t^1 - Z_s^1) &= 5(t - s)\end{aligned}$$

$$E(Z_t^1 - Z_s^1) = E(Y_t - Y_s) + E(t - s) = -1(t - s) + (t - s) = 0$$

The distribution of $(Z_t^1 - Z_s^1)$ is normal with mean 0 and variance $5(t - s)$.

$$\begin{aligned}Z_t^2 - Z_s^2 &= -B_t + 3W_t + B_s - 3W_s \\Z_t^2 - Z_s^2 &= -(B_t - B_s) + 3(W_t - W_s)\end{aligned}$$

The distribution of $(Z_t^2 - Z_s^2)$ depends on the distributions of $(B_t - B_s)$ and $(W_t - W_s)$.
The distributions of $(B_t - B_s)$ and $(W_t - W_s)$ are normal with mean 0 and variance $(t - s)$.

$$\begin{aligned}\text{Var}(Z_t^2 - Z_s^2) &= \text{Var}\{-(B_t - B_s)\} + \text{Var}\{3(W_t - W_s)\} \\ \text{Var}(Z_t^2 - Z_s^2) &= \text{Var}(B_t - B_s) + 9\text{Var}(W_t - W_s) \\ \text{Var}(Z_t^2 - Z_s^2) &= (t - s) + 9(t - s) \\ \text{Var}(Z_t^2 - Z_s^2) &= 10(t - s)\end{aligned}$$

$$E(Z_t^2 - Z_s^2) = -E(B_t - B_s) + 3E(W_t - W_s) = 0 + 3 * 0 = 0$$

The distribution of $(Z_t^2 - Z_s^2)$ is normal with mean 0 and variance $10(t - s)$.

Because the distributions of both $(Z_t^1 - Z_s^1)$ and $(Z_t^2 - Z_s^2)$ are normal with mean 0 and some variance, distribution of $Z_t - Z_s$ is joint normal with mean 0 and covariance matrix $\Gamma(t - s)$. The second property is satisfied.

Property 3

$$Z_t - Z_s = (Z_t^1 - Z_s^1, Z_t^2 - Z_s^2)$$

$$\begin{aligned} Z_t^1 - Z_s^1 &= 2(B_t - B_s) - 1(W_t - W_s) \\ Z_t^2 - Z_s^2 &= -(B_t - B_s) + 3(W_t - W_s) \end{aligned}$$

$(Z_t^1 - Z_s^1)$ and $(Z_t^2 - Z_s^2)$ depends only on the values of $(B_t - B_s)$ and $(W_t - W_s)$. Since B_t and W_t are Brownian motions, $(B_t - B_s)$ and $(W_t - W_s)$ are independent of \mathcal{F}_s .

Since the components are independent of \mathcal{F}_s , $(Z_t^1 - Z_s^1)$ and $(Z_t^2 - Z_s^2)$ are also independent of \mathcal{F}_s .

Since both $(Z_t^1 - Z_s^1)$ and $(Z_t^2 - Z_s^2)$ are independent of \mathcal{F}_s , the random vector $Z_t - Z_s$ is independent of \mathcal{F}_s . The third property is satisfied.

Property 4

B_t and W_t are Brownian motions. They are both continuous functions of t .

Since Z_t^1 and Z_t^2 are merely linear combinations of B_t and W_t , a linear combination of continuous functions of t , is also a continuous function of t . Therefore, with probability one, the functions $t \rightarrow Z_t^1$ and $t \rightarrow Z_t^2$ are continuous.

Since with probability one, the functions $t \rightarrow Z_t^1$ and $t \rightarrow Z_t^2$ are continuous, then with probability one, the function $t \rightarrow Z_t$ is continuous.

Conclusion: Since Z_t satisfies all 4 properties, Z_t is a two-dimensional Brownian motion starting at the origin $\{Z_0 = (0, 0)\}$.

Covariance Matrix

$$\Gamma_t = \begin{pmatrix} \text{Var}(Z_t^1) & \text{Cov}(Z_t^1, Z_t^2) \\ \text{Cov}(Z_t^1, Z_t^2) & \text{Var}(Z_t^2) \end{pmatrix}$$

$$\begin{aligned} Z_t^1 &= Y_t + t \\ \text{Var}(Z_t^1) &= \text{Var}(Y_t) + \text{Var}(t) = 5t \end{aligned}$$

$$\begin{aligned} Z_t^2 &= -B_t + 3W_t \\ \text{Var}(Z_t^2) &= \text{Var}\{-B_t\} + \text{Var}\{3W_t\} = \text{Var}(B_t) + 9\text{Var}(W_t) = t + 9t \\ \text{Var}(Z_t^2) &= 10t \end{aligned}$$

$$\begin{aligned}Z_t^1 &= Y_t + t = 2B_t - W_t - t + t \\Z_t^1 &= 2B_t - W_t \\Z_t^2 &= -B_t + 3W_t\end{aligned}$$

$$\text{Cov}(Z_t^1, Z_t^2) = \text{Cov}(2B_t, -B_t) + \text{Cov}(2B_t, 3W_t) + \text{Cov}(-W_t, -B_t) + \text{Cov}(-W_t, 3W_t)$$

Since B_t and W_t are independent Brownian motions,

$$\text{Cov}(W_t, B_t) = \text{Cov}(2B_t, 3W_t) = \text{Cov}(-W_t, -B_t) = 0$$

$$\begin{aligned}\text{Cov}(B_t, B_t) &= \text{Var}(B_t) = t \\ \text{Cov}(2B_t, -B_t) &= (2) * (-1) * \text{Cov}(B_t, B_t) = -2t\end{aligned}$$

$$\begin{aligned}\text{Cov}(W_t, W_t) &= \text{Var}(W_t) = t \\ \text{Cov}(-W_t, 3W_t) &= (-1) * (3) * \text{Cov}(W_t, W_t) = -3t\end{aligned}$$

$$\text{Cov}(Z_t^1, Z_t^2) = -2t + 0 + 0 - 3t = -5t$$

$$\Gamma_t = \begin{pmatrix} 5t & -5t \\ -5t & 10t \end{pmatrix}$$

Covariance Matrix

$$\Gamma = \begin{pmatrix} 5 & -5 \\ -5 & 10 \end{pmatrix}$$

$$\begin{aligned}E(Z_t^1) &= 2E(B_t) - E(W_t) = 0 - 0 = 0 \\ E(Z_t^2) &= -E(B_t) + 3E(W_t) = 0 + 3 * 0 = 0\end{aligned}$$

Since $E(Z_t^1) = E(Z_t^2) = 0$, Z_t has 0 drift.

Exercise 2.3

According to theorem 2.8.1 in page 66, since Z_t^1 is a Brownian motion with drift m and variance σ^2 ,

$$\langle Z^1 \rangle_t = \sigma^2 t = 5t$$

According to theorem 2.9.1 in page 69, since Z_t is a 2-dimensional Brownian motion with drift m and covariance matrix Γ ,

$$\langle Z^1, Z^2 \rangle_t = \Gamma_{12}t = -5t$$

Exercise 3

$$P(B_t \geq 6W_t - 4) = P(4 \geq 6W_t - B_t) = P(6W_t - B_t \leq 4) = P(X_t \leq 4)$$

$$E(6W_t - B_t) = 6E(W_t) - E(B_t) = 0 - 0 = 0$$

$$\begin{aligned} \text{Var}(6W_t - B_t) &= \text{Var}(6W_t) + \text{Var}(-B_t) \\ &= 36\text{Var}(W_t) + \text{Var}(B_t) = 36t + t = 37t \end{aligned}$$

$$X_t \sim N(0, 37t), \sigma = \sqrt{37}$$

$$X_t = 6W_t - B_t$$

$$= P\left(Z_t \leq \frac{4-0}{\sqrt{37}} = \frac{4}{\sqrt{37}}\right)$$

Using reflection principle,
page 61 class notes

Let Z_t as a standard brownian motion, $Z_t = \frac{X_t}{\sqrt{37}}$

$$\begin{aligned} P(\max_{0 \leq t \leq 3} X_t \geq 4) &= P(\max_{0 \leq t \leq 3} Z_t \geq \frac{4}{\sqrt{37}}) = 2\left[1 - \Phi\left(\frac{4}{\sqrt{37}} / \sqrt{3}\right)\right] \\ &= 2\left[1 - \Phi(0.38)\right] = 2 - 2\Phi(0.38) \end{aligned}$$

$$P(B_t \geq 6W_t - 4 \text{ for all } 0 \leq t \leq 3) = P(\max_{0 \leq t \leq 3} 6W_t - B_t \leq 4)$$

$$= P(\max_{0 \leq t \leq 3} X_t \leq 4) = 1 - P(\max_{0 \leq t \leq 3} X_t \geq 4)$$

$$= 2\Phi(0.38) - 1 = 2(0.64803) - 1$$

$$P(B_t \geq 6W_t - 4 \text{ for all } 0 \leq t \leq 3) = 0.2961 //$$

Exercise 4

For $B_t \sim N(0, t) \rightarrow B_t$ is standard Brownian motion,
the functions $\phi(t, x)$ satisfies the heat equation:

$$\partial_t \phi(t, x) = \frac{1}{2} \partial_{xx} \phi(t, x), \quad \rightarrow \text{PDE}$$

for each $\phi(t, x)$ in exercise 4.1-4.4
(class notes page 71)

If B has drift 1 and variance parameter 4,
the functions $\phi(t, x)$ satisfies the heat equation:

$$\partial_t \phi(t, x) = -1 \partial_x \phi(t, x) + \frac{4}{2} \partial_{xx} \phi(t, x)$$

$$\partial_t \phi(t, x) = -\partial_x \phi(t, x) + 2 \partial_{xx} \phi(t, x) \quad \rightarrow \text{PDE}$$

for each function $\phi(t, x)$ in exercise 4.1-4.4

(class notes page 73)

4.1

$$\phi(t, x) = \frac{1}{\sqrt{2\pi t}} e^{-x^2/2t} \quad (\text{class notes page 70})$$

initial condition: $B_0 = 0$

For 4.3, initial condition:

$$B_0 = B_0 = x$$

$$\phi(0, x) = E[B_0^2 | B_0 = x] \cdot E[x^2] = E[x^2] = x^2$$

$$\text{initial condition: } \phi(0, x) = x^2$$

For 4.4, final condition:

$$B_4 = B_4 = x$$

$$\phi(4, x) = E[B_4 - B_4^2 | B_4 = x] \cdot E[x - x^2] = x - x^2$$

$$\text{Final condition: } \phi(4, x) = x - x^2$$

These conditions apply when B is a standard Brownian motion, and also for when B has drift 1 and variance parameter 4.

Exercise 4 (continued)

For 4.2, initial condition:

at $\bar{t}=0$, if $B_0 = x \geq 4$, $P(B_t < 4 | B_0 = x) = 0$
if $B_0 = x < 4$, $P(B_t < 4 | B_0 = x) = 1$

$$\phi(0, x) = \begin{cases} 1 & , \text{ for } x < 4 \\ 0 & , \text{ for } x \geq 4 \end{cases}$$

↳ initial condition

For 4.3, initial condition:

at $t=0$, $B_t = B_0 = x$

$$\phi(0, x) = E[B_0^3 | B_0 = x] = E[B_0^3] = E(x^3) = x^3$$

initial condition: $\phi(0, x) = x^3$

For 4.4, final condition:

at $t=4$, $B_t = B_4 = x$

$$\phi(4, x) = E[B_4 - B_4^2 | B_4 = x] = E[x - x^2] = x - x^2$$

Final condition: $\phi(4, x) = x - x^2$

These conditions apply when B_t is a standard brownian motion, and also for when B has drift 1 and variance parameter 4.

Exercise 5.1

$$S.1) E(Z_t^1) = E(B_t) + E(W_t) - E(t) = 0 + 0 - t = -t = m_1 \cdot t$$

$$E(Z_t^2) = E(2 \cdot B_t) + (-4)E(W_t) = 0 + 0 = 0 = m_2$$

$$\text{drift for } Z = m = (-1, 0) //$$

$$\text{Cov}(B_t, B_t) = \text{Var}(B_t) = t = \text{Var}(W_t) = \text{Cov}(W_t, W_t)$$

$$\text{Cov}(B_t, 2B_t) = (1)(2) \text{Cov}(B_t, B_t) = 2t$$

$$\text{Cov}(B_t, -4W_t) = \text{Cov}(W_t, 2B_t) = \text{Cov}(W_t, B_t) = 0,$$

because B_t and W_t are independent Brownian motions

$$\text{Cov}(W_t, -4W_t) = (1)(-4) \text{Cov}(W_t, W_t) = -4t$$

$$\begin{aligned} \text{Cov}(Z_t^1, Z_t^2) &= \text{Cov}(B_t, 2B_t) + \text{Cov}(B_t, -4W_t) + \text{Cov}(W_t, 2B_t) \\ &\quad + \text{Cov}(W_t, -4W_t) \\ &= 2t + 0 - 4t = -2t \end{aligned}$$

$$\text{Var}(Z_t^1) = \text{Var}(B_t) + \text{Var}(W_t) = t + t = 2t$$

$$\begin{aligned} \text{Var}(Z_t^2) &= \text{Var}(2B_t) + \text{Var}(-4W_t) = 4\text{Var}(B_t) + 16\text{Var}(W_t) \\ &= 4t + 16t = 20t \end{aligned}$$

Covariance matrix for Z

$$T_t = \begin{pmatrix} \text{Var}(Z_t^1) & \text{Cov}(Z_t^1, Z_t^2) \\ \text{Cov}(Z_t^1, Z_t^2) & \text{Var}(Z_t^2) \end{pmatrix} = \begin{pmatrix} 2t & -2t \\ -2t & 20t \end{pmatrix}$$

$$T = \begin{pmatrix} 2 & -2 \\ -2 & 20 \end{pmatrix} //$$

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Exercise 5.2

According to theorem 2.10.3 (page 76 class notes),

Since z is a 2-dimensional Brownian motion with drift $m=(-1,0)$ and covariance matrix T , then the infinitesimal generator is

$$L f(x) = m \cdot \nabla f(x) + \frac{1}{2} \sum_{j=1}^d \sum_{k=1}^d T_{jk} \frac{\partial^2 f}{\partial x^j \partial x^k}(x)$$

$$m \cdot \nabla f(x) = (-1, 0) \cdot \left(\frac{\partial f}{\partial x^1}, \frac{\partial f}{\partial x^2} \right) = -\frac{\partial f}{\partial x^1}$$

$$\frac{1}{2} \sum_{j=1}^d \sum_{k=1}^d T_{jk} \frac{\partial^2 f}{\partial x^j \partial x^k}(x) = \frac{1}{2} \left\{ 2 \cdot \frac{\partial^2 f}{(\partial x^1)^2} + (-2) \cdot 2 \cdot \frac{\partial^2 f}{\partial x^1 \partial x^2} + 20 \cdot \frac{\partial^2 f}{(\partial x^2)^2} \right\}$$

$$= \frac{\partial^2 f}{(\partial x^1)^2} - 2 \cdot \frac{\partial^2 f}{\partial x^1 \partial x^2} + 10 \cdot \frac{\partial^2 f}{(\partial x^2)^2}$$

$$L f(x) = -\frac{\partial f}{\partial x^1} + \frac{\partial^2 f}{(\partial x^1)^2} - 2 \cdot \frac{\partial^2 f}{\partial x^1 \partial x^2} + 10 \cdot \frac{\partial^2 f}{(\partial x^2)^2} //$$

↳ L operator associated to z

$$L^* f(x) = \frac{\partial f}{\partial x^1} + \frac{\partial^2 f}{(\partial x^1)^2} - 2 \cdot \frac{\partial^2 f}{\partial x^1 \partial x^2} + 10 \cdot \frac{\partial^2 f}{(\partial x^2)^2} //$$

↳ L^* operator associated to z

(page 79 class notes)

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Exercise 5.3

According to page 79 (class notes), $\phi(t, x)$ satisfies the equation $\partial_t \phi(t, x) = L_x^* \phi(t, x)$

$$\partial_t \phi(t, x) = \frac{\partial \phi(x)}{\partial x^1} + \frac{\partial^2 \phi(x)}{\partial (x^1)^2} - 2 \cdot \frac{\partial^2 \phi(x)}{\partial x^1 \partial x^2} + 10 \cdot \frac{\partial^2 \phi(x)}{\partial (x^2)^2}$$

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