

9.1 Motivation

In the previous chapter we analyzed the strong convergence of the stochastic theta method to the solution X of a stochastic differential equation

$$(9.1) \quad \begin{aligned} dX(t) &= f(t, X(t)) dt + g(t, X(t)) dW(t) \\ X(0) &= X_0 \end{aligned}$$

In particular, under suitable conditions on f, g, X_0 there exists a constant $C > 0$ such that

$$(9.2) \quad \left(\mathbb{E} \left[\max_{0 \leq i \leq n} |X_n(t_i) - X(t_i)|^2 \right] \right)^{1/2} \leq C |h|^{1/2} \quad \forall h \text{ with } |h| \leq \bar{h},$$

where X_n denotes the grid fct. generated by the stochastic theta method. From (9.2) it follows that X_n gives a good pathwise approximation of the exact solution.

However, in many applications we are only interested in approximating an expected value of some quantity of $X(T)$, that is

$$\mathbb{E}[\varphi(X(T))]$$

for a fct. $\varphi: \mathbb{R}^m \rightarrow \mathbb{R}$.

Example 9.1

In the Black-Scholes model we encountered the problem to give an approximation of the discounted expected payoff

$$\mathbb{E}[e^{-rT} \max(0, S(T))], \quad (\varphi(x) = e^{-rT} \max(0, x))$$

where $S(t)$ is the solution to the SODE

$$dS(t) = rS(t)dt + \sigma S(t)dW(t), \quad S(0) = S_0.$$

In this chapter we will work with following assumptions.

~~(P1)~~ $\varphi \in C_p^2(\mathbb{R})$, that is $\varphi \in C^2(\mathbb{R})$ and $\exists C, q > 0$ such that $|\varphi^{(i)}(x)| \leq C(1+|x|^q) \quad \forall x \in \mathbb{R}$
^ polynomial growth

Further, we only consider scalar SODEs ($m=d=1$) with the usual Lipschitz- and linear growth conditions on $f: [0, T] \times \mathbb{R} \rightarrow \mathbb{R}, g: [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$.

Our aim is to prove, that $\forall \varphi \in C_p^2(\mathbb{R}) \exists C(\varphi) > 0$ s.t.
(9.3) $|\mathbb{E}[\varphi(X(T)) - \varphi(X_n(T))]| \leq C(\varphi) |h|^\gamma$.

We refer to (9.3) as the weak error of convergence of the stochastic theta method (with order $\gamma = 1$)

Remark 9.2

From (9.2) it directly follows the weak convergence with order $\gamma = \frac{1}{2}$:

$$|\mathbb{E}[\varphi(X(T)) - \varphi(X_n(T))]| = \underbrace{|\mathbb{E}[\int_0^1 \varphi'(X_n(T) + s(X(T) - X_n(T))) ds (X(T) - X_n(T))] |}_{\text{Taylor}} \leq \underbrace{\left(\mathbb{E} \left[\left(\int_0^1 \varphi'(X_n(T) + s(X(T) - X_n(T))) ds \right)^2 \right] \right)^{1/2}}_{\substack{\uparrow \\ \text{Cauchy-Schwarz}}} \underbrace{\left(\mathbb{E}[|X(T) - X_n(T)|^2] \right)^{1/2}}_{\substack{\leq C|h|^{1/2} \\ \uparrow \text{strong error}}}$$

Need to estimate

$$\mathbb{E} \left[\left(\int_0^1 \varphi'(X_n(T) + s(X(T) - X_n(T))) ds \right)^2 \right] \leq \mathbb{E} \left[\int_0^1 C(\varphi) (1 + |X_n(T)|^{2q} + |X(T)|^{2q}) ds \right] \leq \mathbb{E} \left[\int_0^1 C(\varphi) (1 + |X_n(T)|^{2q} + |X(T)|^{2q}) ds \right]$$

↑ Jensen *polynomial growth, $0 < \alpha < 1$*

$$\leq C(\varphi) (1 + \mathbb{E}[|X_n(T)|^{2q}] + \mathbb{E}[|X(T)|^{2q}])$$

That $\mathbb{E}[|X(T)|^{2q}] < \infty$ is shown as in the estimate in Chapter 7 but we need the additional assumption $\mathbb{E}[|X_0|^{2q}] < \infty$.

Similar for the term $\mathbb{E}[|X_n(T)|^{2q}]$ (see at the end). (for example $X_0 = x_0 \in \mathbb{R}$ constant random variable)

□

9.2 Kolmogorov backward equation

For the analysis of the weak error of convergence we will make use of a link between SODEs and PDEs

For this we consider the PDE (Kolmogorov backward equation)

$$(9.4) \quad \begin{cases} \frac{\partial}{\partial t} u(t,x) + \frac{1}{2} a(t,x) \frac{\partial^2}{\partial x^2} u(t,x) + f(t,x) \frac{\partial}{\partial x} u(t,x) = 0, & x \in \mathbb{R}, t \in [0, T] \\ u(T, x) = \varphi(x) \end{cases}$$

where $a: [0, T] \times \mathbb{R} \rightarrow [0, \infty)$, $f: [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$, $\varphi: \mathbb{R} \rightarrow \mathbb{R}$ are continuous.

Define

$$g(t, x) = \sqrt{a(t, x)}$$

and assume further that f, g satisfy a global Lipschitz condition. Then the SODE

$$(9.5) \quad \begin{cases} dz(s) = f(s, z(s)) ds + g(s, X(s)) dW(s) & \text{on } s \in [t, T] \\ z(t) = x \end{cases}$$

has a unique solution, which we denote by $Z(s; t, x)$.

The link between (9.4) and (9.5) is given by the following special case of the Feynman-Kac formula:

Theorem 9.3

9.4

Assume that there exists a smooth solution u to (9.4), that is $u \in C^{2,1}(\mathbb{R} \times [0, T] \times \mathbb{R})$ and $u(t, \cdot) \in C_p^2(\mathbb{R})$ (uniformly in t)

Then for all $t \leq s \leq T$

$$u(t, x) = \mathbb{E}[u(s, z(s; t, x))] \quad \forall x \in \mathbb{R}$$

In particular

$$(9.6) \quad u(t, x) = \mathbb{E}[\psi(z(T; t, x))] \quad \forall x \in \mathbb{R}$$

Proof:

Apply Itô's formula to $u(s, z(s; t, x)) =: u(s, z(s))$

$$du(s, z(s)) = \underbrace{\left(\frac{\partial}{\partial t} u(s, z(s)) + \frac{\partial}{\partial x} u(s, z(s)) f(s, z(s)) + \frac{1}{2} \frac{\partial^2}{\partial x^2} u(s, z(s)) g^2(s, z(s)) \right)}_{\equiv 0 \quad \forall s \text{ (9.4)}} ds + \frac{\partial}{\partial x} u(s, z(s)) g(s, z(s)) dW(s)$$

or in integral form

$$u(T, z(T)) - u(s, z(s)) = \int_s^T \frac{\partial}{\partial x} u(\eta, z(\eta)) g(\eta, z(\eta)) dW(\eta)$$

Well-defined using polynomial growth of e_x , linear growth of g and existence of all moments of $z(s)$!

Taking expectation: $\forall t \leq s \leq T$:

$$\Rightarrow \mathbb{E}[u(T, z(T))] = \mathbb{E}[u(s, z(s))]$$

$$\mathbb{E}[\psi(z(T; t, x))]$$

$$\left(\begin{aligned} &= \mathbb{E}[u(t, z(t; t, x))] = u(t, x) \\ & \text{by } x \end{aligned} \right)$$

(Start here with lecture on Friday)

□

Example 9.4

Feynman-Kac formula gives explicit representation of solutions to PDEs.

Consider the heat equation

$$\begin{cases} \frac{\partial v}{\partial t} = \frac{1}{2} \frac{\partial^2 v}{\partial x^2}, & \text{on } [0, T] \times \mathbb{R} \\ v(0, x) = \varphi(x) & \forall x \in \mathbb{R} \end{cases} \quad (9.7)$$

Transform (9.7) into a backward-equation by setting $u(t, x) = v(T-t, x)$ ($u(T-t, x) = v(\bar{t}, x)$)

Then
$$\frac{\partial}{\partial t} u(t, x) = (-1) \frac{\partial}{\partial t} v(T-t, x) = (-1) \frac{\partial^2}{\partial x^2} v(T-t, x) \stackrel{(9.7)}{=} (-1) \frac{\partial^2}{\partial x^2} u(t, x)$$

that is, u solves

$$\begin{cases} \frac{\partial}{\partial t} u + \frac{1}{2} \frac{\partial^2}{\partial x^2} u = 0 \\ u(T, x) = \varphi(x) \end{cases}$$

The corresponding SODE is

$$dZ(s) = 1 dW(s), \quad Z(t) = x$$

with solution $Z(s; t, x) = W(s) - W(t) + x$.

Hence

$$\Rightarrow \text{Th. 9.3} \quad u(t, x) = \mathbb{E}[\varphi(Z(T; t, x))] = \mathbb{E}[\varphi(\underbrace{W(T) - W(t)}_{\sim N(0, T-t)} + x)]$$

$$\begin{aligned} &= \int_{\mathbb{R}} \varphi(y+x) f_{0, T-t}(y) dy \\ &= \frac{1}{\sqrt{2\pi(T-t)}} \int_{\mathbb{R}} \varphi(y+x) e^{-\frac{1}{2} \frac{y^2}{(T-t)}} dy \\ &= \frac{1}{\sqrt{2\pi(T-t)}} \int_{\mathbb{R}} \varphi(y) e^{-\frac{1}{2} \frac{(x-y)^2}{(T-t)}} dy \end{aligned}$$

$$\text{or } v(t, x) = u(T-t, x) = \frac{1}{\sqrt{2\pi t}} \int_{\mathbb{R}} \varphi(y) e^{-\frac{(x-y)^2}{2t}} dy$$

□

Example 9.5 (perhaps skip this example)

Consider the geo. B.M.

$$dS(t) = \mu S(t) dt + \sigma S(t) dW(t), \quad S(0) = S_0,$$

where $S_0 \in [0, \infty)$ is deterministic, $\mu, \sigma > 0$,

We want to approximate $\mathbb{E}[\varphi(S(T))]$ for some $\varphi: \mathbb{R} \rightarrow \mathbb{R}$

The corresponding K.B. Eq is

$$(9.8) \quad \begin{cases} \frac{\partial}{\partial t} u(t, x) + \frac{1}{2} \sigma^2 x^2 \frac{\partial^2}{\partial x^2} u(t, x) + \mu x \frac{\partial}{\partial x} u(t, x) = 0 \\ u(T, x) = \varphi(x) \end{cases}$$

Since we do not know if there exists a solution to (9.8) we apply the ansatz:

$$(9.9) \quad \hat{u}(t, x) = \mathbb{E}[\varphi(S(T; t, x))]$$

Since (general initial data):

$$S(T; t, x) = x \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + \sigma(W(T) - W(t))\right)$$

we get

$$\hat{u}(t, x) = \mathbb{E}\left[\varphi\left(x \cdot \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + \sigma(W(T) - W(t))\right)\right)\right]$$

$$= \int_{\mathbb{R}} \varphi\left(x \cdot \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + \sigma\sqrt{T-t}y\right)\right) f_{0,1}(y) dy$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} \varphi\left(x \cdot \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + \sigma\sqrt{T-t}y\right)\right) \exp\left(-\frac{1}{2}y^2\right) dy$$

As one can see, $\hat{u}(t, x)$ is differentiable w.r.t (t, x) :

$$\frac{\partial}{\partial t} \hat{u}(t, x) = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} \varphi'\left(x \cdot \exp\left(\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + \sigma\sqrt{T-t}y\right)\right) \cdot \left(-\left(\mu - \frac{1}{2}\sigma^2\right) + \frac{1}{2}\sigma y \cdot \frac{1}{\sqrt{T-t}}\right) \exp\left(-\frac{1}{2}y^2\right) dy$$

and similar for $\frac{\partial}{\partial x}, \frac{\partial^2}{\partial x^2}$. It is an exercise to verify that

$\hat{u}(t, x)$ is a solution to (9.8).

(Example 9.5 continued)

9.7

Remember that we want to approximate

$$\mathbb{E}[\varphi(S(T; t, x))]$$

By using (9.9) we can do this by approximating (9.8) using numerical methods for deterministic PDEs.

The following theorem (which we quote from Friedman [1975]) gives sufficient conditions on f, g , such that the K.B. Eq has a smooth solution. □

Theorem 9.6

Let $f, g \in C([0, T] \times \mathbb{R})$, such that $f(t, \cdot), g(t, \cdot) \in C^2(\mathbb{R})$ with bounded derivatives and $\varphi \in C_p^2(\mathbb{R})$.

Then the function (Z is the solution to (9.5))

$$u(t, x) = \mathbb{E}[\varphi(Z(T; t, x))]$$

has continuous derivatives $\frac{\partial u}{\partial t}, \frac{\partial u}{\partial x}, \frac{\partial^2 u}{\partial x^2}$ in $[0, T] \times \mathbb{R}$

and is a solution to (9.4).

Moreover, $\exists C, \rho$ independent of t such that

$$\left| \frac{\partial}{\partial x} u(t, x) \right| + \left| \frac{\partial^2}{\partial x^2} u(t, x) \right| \leq C(1 + |x|^\rho).$$

(The proof relies on differentiation of $Z(s; t, x)$ w.r.t x)

9.3 Weak convergence of Euler-Maruyama

9.8

As above we consider the SODE

$$dX(t) = f(t, X(t)) dt + g(t, X(t)) dW(t) \quad (9.10)$$

$$X(0) = x_0$$

with

$x_0 \in \mathbb{R}$, f, g as in Th. 9.6

Then the corresponding K.B. Eq

$$\begin{cases} \frac{\partial}{\partial t} u(t, x) + \frac{1}{2} g^2(t, x) \frac{\partial^2}{\partial x^2} u(t, x) + f(t, x) \frac{\partial}{\partial x} u(t, x) = 0 \\ u(T, x) = \varphi(x) \end{cases} \quad (9.11)$$

has a smooth solution for every $\varphi \in C_p^2(\mathbb{R})$

Theorem 9.7 In addition to the assumptions of Th. 9.6 let the

mappings $\begin{cases} (t, x) \mapsto u_{xx}(t, x) (g^2(t, x) - g^2(t, x)) \\ (t, x) \mapsto u_x(t, x) (f(s, y) - f(t, x)) \end{cases}$ be sufficiently "smooth"

such that all derivatives which appear in the proof have polynomial growth.

Proof: Then the Euler-Maruyama method is weakly convergent with order $\gamma=1$.

By Th. 9.3 we have

$$u(0, x_0) = \mathbb{E}[\varphi(X(T))] \quad (9.12)$$

where u solves (9.11).

Next, we interpolate the grid functions X_h to a continuous

Itô-process by setting

$$X_h(t) = x_0 + \int_0^t \bar{f}(s) ds + \int_0^t \bar{g}(s) dW(s)$$

where

$$\bar{f}(s) = \sum_{i=1}^{M_h} \mathbb{1}_{[t_{i-1}, t_i]}(s) f(t_{i-1}, X_h(t_{i-1})), \quad \bar{g}(s) = \sum_{i=1}^{M_h} \mathbb{1}_{[t_{i-1}, t_i]}(s) g(t_{i-1}, X_h(t_{i-1}))$$

An application of Itô's formula yields

$$\begin{aligned}
 & u(T, X_n(T)) - u(0, X_n(0)) \\
 &= \int_0^T u_t(s, X_n(s)) + u_x(s, X_n(s)) \bar{f}(s) + \frac{1}{2} u_{xx}(s, X_n(s)) \bar{g}^2(s) ds \\
 &+ \int_0^T u_x(s, X_n(s)) \bar{g}(s) dW(s).
 \end{aligned}$$

(9.11) gives us the relation

$$u_t(s, X_n(s)) = -f(s, X_n(s)) \frac{\partial}{\partial x} u(s, X_n(s)) - \frac{1}{2} g^2(s, X_n(s)) \frac{\partial^2}{\partial x^2} u(s, X_n(s)).$$

Hence

$$\begin{aligned}
 u(T, X_n(T)) - u(0, \underbrace{X_n(0)}_{X_0}) &= \int_0^T u_x(s, X_n(s)) (\bar{f}(s) - f(s, X_n(s))) + \frac{1}{2} \frac{\partial^2}{\partial x^2} u(s, X_n(s)) (\bar{g}^2(s) - g^2(s, X_n(s))) ds \\
 &+ \int_0^T u_x(s, X_n(s)) \bar{g}(s) dW(s)
 \end{aligned}$$

Therefore

$$\begin{aligned}
 & \mathbb{E}[u(X_n(T)) - u(X(T))] \quad X_n = X_n(0) = u(0, X_0) \\
 & \stackrel{(9.12)}{=} \mathbb{E}[u(T, X_n(T)) - u(0, X_0)] \\
 &= \int_0^T \mathbb{E}\left[u_x(s, X_n(s)) (\bar{f}(s) - f(s, X_n(s))) + \frac{1}{2} \frac{\partial^2}{\partial x^2} u(s, X_n(s)) (\bar{g}^2(s) - g^2(s, X_n(s))) \right] ds \\
 &= \sum_{i=1}^{M_n} \int_{t_{i-1}}^{t_i} \mathbb{E}\left[u_x(s, X_n(s)) (f(t_{i-1}, X_n(t_{i-1})) - f(s, X_n(s))) \right] ds \\
 &+ \sum_{i=1}^{M_n} \frac{1}{2} \int_{t_{i-1}}^{t_i} \mathbb{E}\left[u_{xx}(s, X_n(s)) (g^2(t_{i-1}, X_n(t_{i-1})) - g^2(s, X_n(s))) \right] ds
 \end{aligned}$$

It remains to show

$$(9.13) \quad \sup_{S \in (t_{i-1}, t_i)} \left| \mathbb{E}\left[u_{xx}(s, X_n(s)) (g^2(t_{i-1}, X_n(t_{i-1})) - g^2(s, X_n(s))) \right] \right| \leq C |h|$$

$$(9.14) \quad \text{and} \quad \sup_{S \in (t_{i-1}, t_i)} \left| \mathbb{E}\left[u_{x_0}(s, X_n(s)) (f(t_{i-1}, X_n(t_{i-1})) - f(s, X_n(s))) \right] \right| \leq C |h|$$

↑
indep of i

and

$$(9.15) \sup_{s \in (t_{i-1}, t_i)} (1-\theta) | \mathbb{E} [u_x(s, X_h(s)) (f(t_{i-1}, X_h(t_{i-1})) - f(s, X_h(s)))] | \leq Ch$$

If this is established we have shown

$$| \mathbb{E} [\varphi(X_h(T)) - \varphi(T)] | \leq CT |h|$$

which is the weak convergence with order $\gamma=1$.

We only demonstrated the estimate (9.13).

For this define for $t \in [t_{i-1}, t_i]$

$$v(t, x, y) = u_{xx}(t, x) (g^2(t_{i-1}, y) - g^2(t, x))$$

Note that (9.13) has the form

$$| \mathbb{E} [v(s, X_h(s), X_h(t_{i-1}))] |$$

In particular $v(t_{i-1}, X_h(t_{i-1}), X_h(t_{i-1})) = 0$.

Apply Itô's formula to $t \mapsto v(t, X_h(t), y)$, $y \in \mathbb{R}$ fixed, to get

$$v(s, X_h(s), y) = v(t_{i-1}, X_h(t_{i-1}), y) = \int_{t_{i-1}}^s (v_t + v_x \bar{f} + \frac{1}{2} v_{xx} \bar{g}^2) d\eta + \int_{t_{i-1}}^s v_x \bar{g} dW(\eta)$$

\searrow vanishes!

$$\Rightarrow | \mathbb{E} [v(s, X_h(s), X_h(t_{i-1}))] | = \left| \int_{t_{i-1}}^s \mathbb{E} [v_t(\eta, X_h(\eta), X_h(t_{i-1})) + v_x(\dots) \bar{f}(\eta) + \frac{1}{2} v_{xx}(\dots) \bar{g}^2(\eta)] d\eta \right|$$

$$\leq \int_{t_{i-1}}^{t_i} \mathbb{E} [|v_t| + |v_x(\dots) \bar{f}(\eta)| + \frac{1}{2} |v_{xx}(\dots) \bar{g}^2(\eta)|] d\eta$$

$f(t_{i-1}, X_h(t_{i-1}))$ $= g^2(t_{i-1}, X_h(t_{i-1}))$

$$\leq Ch_i (1 + \mathbb{E} [\max_{0 \leq i \leq M_h} |X_h(t_i)|^q]) \leq C |h|$$

\uparrow polynomial growth of v_t, v_x etc.

$< +\infty$

□

Lemma 9.8

Assume $\mathbb{E}[|X_0|^{2q}] < \infty$ for $q \geq 1$. Then $\exists C$ and $\bar{h} > 0$

such that $\mathbb{E}[\sup_{0 \leq i \leq M_h} |X_h(t_i)|^{2q}] \leq C \quad \forall h$ with $|h| < \bar{h}$.

Proof

$$\left(\mathbb{E} \left[\sup_{0 \leq i \leq M_h} |X_h(t_i)|^{2q} \right] \right)^{1/2q} = \|X_h\|_{2q, h, 0}$$

$$\leq C_1 \|L_h(X_h)\|_{2q, h, -1} = C_1 \left(\|X_0\|_{L^{2q}(\mathbb{R})} + \left(\mathbb{E} \left[\max_{1 \leq j \leq M_h} \left| \sum_{i=1}^j X_h(t_i) - X_h(t_{i-1}) \right|^{2q} \right] \right)^{1/2q} \right)$$

L_h is stable

and

$$\left(\mathbb{E} \left[\max_{1 \leq j \leq \nu} \left| \sum_{i=1}^j X_h(t_i) - X_h(t_{i-1}) \right|^{2q} \right] \right)^{1/2q}$$

$$\leq \left(\mathbb{E} \left[\max_{1 \leq j \leq M_h} \left| \sum_{i=1}^j (1-\theta) h_i f(t_i, X_h(t_{i-1})) \right|^{2q} \right] \right)^{1/2q}$$

$$+ \left(\mathbb{E} \left[\max_{1 \leq j \leq M_h} \left| \sum_{i=1}^j \theta h_i f(t_i, X_h(t_i)) \right|^{2q} \right] \right)^{1/2q}$$

$$+ \left(\mathbb{E} \left[\max_{1 \leq j \leq \nu} \left| \sum_{i=1}^j g(t_i, X_h(t_{i-1})) \Delta v_i \right|^{2q} \right] \right)^{1/2q}$$

$$\leq T^{\frac{2q-1}{2q}} \left(\mathbb{E} \left[\max_{1 \leq j \leq M_h} \sum_{i=1}^j (1-\theta)^{2q} h_i^{2q} (1 + |X_h(t_{i-1})|^{2q}) \right] \right)^{1/2q}$$

$$+ T^{\frac{2q-1}{2q}} \left(\mathbb{E} \left[\max_{1 \leq j \leq M_h} \sum_{i=1}^j \theta^{2q} h_i^{2q} (1 + |X_h(t_i)|^{2q}) \right] \right)^{1/2q}$$

$$+ \left((q) T^{\frac{2q-2}{2q}} \left(\int_0^{\nu} \mathbb{E} \left[\left| \sum_{i=1}^{\nu} \mathbb{1}_{[t_{i-1}, t_i)}(s) g(t_{i-1}, X_h(t_{i-1})) \right|^{2q} \right] ds \right) \right)^{1/2q}$$

$$\leq \sum_{i=1}^{M_h} h_i^{2q} (1 + |X_h(t_{i-1})|^{2q})$$

$$\leq L T^{\frac{2q-1}{2q}} (1 + (q)) \left(\sum_{i=1}^{\nu} h_i \mathbb{E} \left[|X_h(t_i)|^{2q} \right] \right)^{1/2q} \leq L T^{\frac{2q-1}{2q}} (1 + (q)) \left(\sum_{i=1}^{\nu} h_i \mathbb{E} \left[\max_{0 \leq i \leq \nu} |X_h(t_i)|^{2q} \right] \right)^{1/2q}$$

Then take $()^{2q}$, bring highest summand on other side, apply discrete Gronwall. \square

For example 9.5 (Solution to the exercise)

Appendix 9.7

$$\frac{d}{dt} \varphi(x \cdot \exp((\mu - \frac{1}{2}\sigma^2)(T-t) + \sigma\sqrt{T-t}y))$$

$$= \varphi'(\dots) \cdot x \cdot \exp((\mu - \frac{1}{2}\sigma^2)(T-t) + \sigma\sqrt{T-t}y) \cdot \left((-\mu + \frac{1}{2}\sigma^2) - \frac{\sigma y}{2\sqrt{T-t}} \right)$$

(Similar for $\frac{\partial}{\partial x}$, $\frac{\partial^2}{\partial x^2}$...)

$\hat{u}(t,x) =$

$$\frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} \varphi(x \cdot \exp((\mu - \frac{1}{2}\sigma^2)(T-t) + \sigma\sqrt{T-t}y)) \exp(-\frac{1}{2}y^2) dy$$

$$\frac{\partial}{\partial x} \varphi(\dots) = \varphi'(\dots) \cdot \exp(\dots)$$

$$\frac{\partial}{\partial y} \varphi(\dots) = \varphi'(\dots) \cdot \sigma\sqrt{T-t} \exp(\dots)$$

$$\frac{\partial^2}{\partial x^2} \varphi(\dots) = \varphi''(\dots) \cdot (\exp(\dots))^2$$

For the second term use integration by parts to obtain

$$\int_{\mathbb{R}} \frac{\partial}{\partial x^2} \varphi(\dots) \exp(-\frac{y^2}{2}) dy = \int_{\mathbb{R}} \varphi''(\dots) \exp(\dots) \cdot \exp(\dots) \exp(-\frac{y^2}{2}) dy$$

$$= 0 - \int_{\mathbb{R}} \varphi'(\dots) \frac{1}{\sigma\sqrt{T-t}} \left(\exp(\dots) (-y) \exp(-\frac{y^2}{2}) \right) dy$$

$$= \int_{\mathbb{R}} \varphi'(\dots) \frac{1}{\sigma\sqrt{T-t}} \left(\sigma\sqrt{T-t} \exp(\dots) \exp(-\frac{y^2}{2}) \right) dy$$

Plug this into (9.8) to obtain:

$$\hat{u}(t,x) = \mathbb{E}[\varphi(S(T; t, x))] \quad (\text{is a strong smooth}$$

solution to (9.8))

□