# Convergence Rates of Random Walk Approximations of Forward-Backward SDEs 

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joint work with
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## Forward Backward Stochastic Differential Equations (FBSDEs)

$$
\begin{aligned}
& X_{t}=x+\int_{0}^{t} b\left(s, X_{s}\right) d s+\int_{0}^{t} \sigma\left(s, X_{s}\right) d B_{s}, \quad 0 \leq t \leq T \\
& Y_{t}=g\left(X_{T}\right)+\int_{t}^{T} f\left(s, X_{s}, Y_{s}, Z_{s}\right) d s-\int_{t}^{T} Z_{s} d B_{s}
\end{aligned}
$$

Replacing $\left(B_{t}\right)_{t \in[0, T]}$ by a random walk $\left(B_{t}^{n}\right)_{t \in[0, T]}$ what kind of convergence one can expect:

$$
\left(\left(Y_{t}^{n}\right)_{t \in[0, T]},\left(Z_{t}^{n}\right)_{t \in[0, T]}\right) \rightarrow\left(\left(Y_{t}\right)_{t \in[0, T]},\left(Z_{t}\right)_{t \in[0, T]}\right) ?
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$$

## Briand, Delyon and Mémin (2001)

If $b, \sigma, f$ and $g$ are Lipschitz and $\left(B_{t}^{n}\right)_{t \in[0, T]}$ such that $\sup _{0 \leq t \leq T}\left|B_{t}^{n}-B_{t}\right| \rightarrow 0, \quad n \rightarrow \infty, \quad$ in probability, then

$$
\sup _{0 \leq t \leq T}\left|Y_{t}^{n}-Y_{t}\right|^{2}+\int_{0}^{T}\left|Z_{s}^{n}-Z_{s}\right|^{2} d s \rightarrow 0 \text { when } n \rightarrow \infty \text { in probability. }
$$

Random walk approximation of the Brownian motion
Let $t_{k}:=k h, k=0, \ldots, n$ be a regular grid of $[0, T]$, where $h=\frac{T}{n}$ and define

$$
\begin{aligned}
& B_{t}^{n}:=\sqrt{h} \sum_{j=1}^{[t / h]} \varepsilon_{j},\left(\varepsilon_{j}\right)_{j=1}^{n} \text { i.i.d. Rademacher r.v.: } \mathbb{P}\left(\varepsilon_{j}= \pm 1\right)=\frac{1}{2} \\
& {\left[B^{n}\right]_{t}=h \sum_{j=1}^{n} j \mathbf{1}_{\left(t_{j-1}, t_{j}\right]}(t) \quad \text { quadratic variation. }}
\end{aligned}
$$

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- Donsker's Theorem:

Convergence of the processes in the Skorokhod space $D[0, T]$ :

$$
\left(B_{t}^{n}\right)_{t \in[0, T]} \rightarrow\left(B_{t}\right)_{t \in[0, T]} \quad \text { in distribution. }
$$

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- We get the $\mathrm{FBS} \Delta \mathrm{E}$ (Forward-backward stochastic difference equation)

$$
\begin{aligned}
& X_{t}^{n}=x+\int_{(0, t]} b\left(s, X_{s-}^{n}\right) d\left[B^{n}\right]_{s}+\int_{(0, t]} \sigma\left(s, X_{s-}^{n}\right) d B_{s}^{n}, \\
& Y_{t}^{n}=g\left(X_{T}^{n}\right)+\int_{(t, T]} f\left(s, X_{s-}^{n} Y_{s-}^{n}, Z_{s}^{n}\right) d\left[B^{n}\right]_{s}-\int_{(t, T]} Z_{s}^{n} d B_{s}^{n}, \quad 0 \leq t \leq T .
\end{aligned}
$$

## BSDEs and $\mathrm{BS} \Delta \mathrm{Es}$

$$
Y_{t}=g\left(B_{s_{1}}, . ., B_{s_{K}}\right)+\int_{t}^{T} f\left(s, Y_{s}, Z_{s}\right) d s-\int_{t}^{T} Z_{s} d B_{s}
$$

$B d$-dimensional $d \geq 2, \Longrightarrow\left(B_{t}^{n}\right)$ does not possess the representation property. ( $M_{t}^{n}$ ) is a martingale orthogonal to $\left(B_{t}^{n}\right)$ :
$Y_{t}^{n}=g\left(B_{s_{1}}^{n}, . ., B_{s_{K}}^{n}\right)+\int_{(t, T]} f\left(s, Y_{s^{-}}^{n}, Z_{s}^{n}\right) d\left[B^{n}\right]_{s}-\int_{(t, T]} Z_{s}^{n} d B_{s}^{n}-\left(M_{T}^{n}-M_{t}^{n}\right)$
Cheridito and Stadje (2013): BSDEs and BS $\Delta$ Es
$f$ sub-quadratic growth in $z$, Lipschitz in $y, g$ bounded and Lipschitz and $\left(B_{t}^{n}\right)_{t \in[0, T]}$ such that
$\mathbb{E}\left[\sup _{0 \leq t \leq T}\left|B_{t}^{n}-B_{t}\right|^{2}\right] \rightarrow 0, \quad n \rightarrow \infty$, then
$\mathbb{E}\left[\sup _{0 \leq t \leq T}\left(\left|Y_{t}^{n}-Y_{t}\right|+\left|\int_{0}^{t} Z_{s}^{n} d B_{s}^{n}-\int_{0}^{t} Z_{s} d B_{s}\right|+\left|M_{t}^{n}\right|\right)^{2}\right] \rightarrow 0$ when $n \rightarrow \infty$.

## Other results

Random walk schemes: convergence in probability or weak convergence:

- Nakayama (2002) (multidimensional), Toldo (2005) (with random terminal time), Numerical schemes: Ma, Protter, San Martín and Torres (2002) (path-dependent terminal condition), Peng, Xu (2008) (Implicit and explicit schemes for BSDEs) Mémin, Peng and Xu (2008), Martinez, San Martín and Torres (2011) (reflected BSDEs), Jańczak $(2008,2009)$ (generalized reflected BSDEs with random terminal time), ...


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Time discretization schemes with $L_{2}$ or $L_{p}$-rate ( $p \geq 2$ )
- Zhang (2004) Bouchard \& Touzi (2004) ,..., Richou (2011), Lionnet \& dos Reis \& Szpruch (2016), S. Geiss \& Ylinen (2018) (regularity of Y), Han \& Jentzen (2017), Chassagneux, Richou (2019), Sun et al. (2022),...


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Random walk schemes: with $L_{2}$-rate

- C.G., Labart, Luoto (2020, 2021) $f$ Lipschitz, $g$ - -Hölder continuous:

$$
\sup _{0 \leq t<T}\left(\mathbb{E}\left|Y_{t}-Y_{t}^{n}\right|^{2}\right)^{\frac{1}{2}} \leq C n^{-\frac{\varepsilon}{4}},\left(\mathbb{E} \int_{0}^{T}\left|Z_{t}-Z_{t}^{n}\right|^{2} d t\right)^{\frac{1}{2}} \leq C_{2} n^{-\beta} \text { for } \beta \in\left(0, \frac{\varepsilon}{4}\right)
$$

## Why only $n^{-\frac{\varepsilon}{4}}$ ?

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$$

- $B^{n}$ is constructed from $B$ by Skorohod embedding:

$$
\left(\mathbb{E}\left|B_{T}^{n}-B_{T}\right|^{2}\right)^{\frac{1}{2}} \leq C n^{-\frac{1}{4}}
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$\Longrightarrow$ the rate for $\left(Y^{n}, Z^{n}\right) \rightarrow(Y, Z)$ can not be expected to be better.

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- Changing the metric to improve the rate?


## Wasserstein distances

For $r>0$ put $\psi_{r}(x):=e^{|x|^{r}}-1$. For a real random variable $X$ we define the Orlicz norm

$$
\|X\|_{\psi_{r}}:=\inf \left\{a>0: \mathbb{E}\left[\psi_{r}(X / a)\right] \leq 1\right\}, \quad \inf \emptyset:=+\infty
$$

Then for any $p>0$,

$$
\|X\|_{\mathrm{L}^{p}} \leq\left(\sup _{x>0}\left\{\frac{x^{p \vee r}}{\psi_{r}(x)}\right\}\right)^{1 /(p \vee r)}\|X\|_{\psi_{r}}
$$

For $X, X^{\prime}$ random variables with $\operatorname{law}(X)=\mu, \operatorname{law}\left(X^{\prime}\right)=\nu$ and $r \geq 1$,

$$
W_{\psi_{r}}(\mu, \nu)=W_{\psi_{r}}\left(X, X^{\prime}\right):=\inf \left\{\left\|Y-Y^{\prime}\right\|_{\psi_{r}}: \operatorname{law}(Y)=\mu, \operatorname{law}\left(Y^{\prime}\right)=\nu\right\} .
$$

is a metric, the Wasserstein distance associated to $\psi_{r}$.

## Wasserstein convergence rates for $B^{n} \rightarrow B$

Theorem 1 (Rio (2009))
$\left(X_{k}\right)_{k \geq 1}$ i.i.d. with
$\mathbb{E}\left[X_{1}\right]=0$,
$\mathbb{E}\left[X_{1}^{2}\right]=1$, and
$\mathbb{E}\left[e^{c\left|X_{1}\right|}\right]<+\infty$ for some $c>0$.
Let $\mathcal{G} \sim N(0,1)$.
Then $\exists C>0$ such that, for $n \geq 1$,

$$
W_{\psi_{1}}\left(\frac{X_{1}+\ldots+X_{n}}{n^{1 / 2}}, \mathcal{G}\right) \leq C n^{-1 / 2} .
$$

For $x \in \mathbb{R}$ and $0 \leq t \leq s \leq T$ we put

$$
B_{s}^{t, x}:=x+B_{s}-B_{t} \quad B_{s}^{n, t, x}:=x+B_{s}^{n}-B_{t}^{n} .
$$

Lemma 2 (Briand, C.G., S.Geiss, Labart (2021))
(1) $\exists C>0$ such that $\forall x \in \mathbb{R}$ and $0 \leq t \leq s \leq T$,

$$
W_{\psi_{1}}\left(B_{s}^{n, t, x}, B_{s}^{t, x}\right) \leq C \sqrt{T} n^{-1 / 2}
$$

(2) If $g: \mathbb{R} \longrightarrow \mathbb{R}$ is $\varepsilon$-Hölder continuous ( $0<\varepsilon \leq 1$ ), then $\forall x \in \mathbb{R}$ and $0 \leq t \leq s \leq T$,

$$
\left|\mathbb{E}\left[g\left(B_{s}^{n, t, x}\right)\right]-\mathbb{E}\left[g\left(B_{s}^{t, x}\right)\right]\right| \leq C\|g\|_{\varepsilon} n^{-\varepsilon / 2},
$$

for some $C=C(T)$.

$$
Y_{t}=g\left(B_{T}\right)+\int_{t}^{T} f\left(s, B_{s}, Y_{s}, Z_{s}\right) d s-\int_{t}^{T} Z_{s} d B_{s}
$$

## Assumption 1

$\exists \varepsilon, \alpha \in(0,1]:$
(1) $g: \mathbb{R} \longrightarrow \mathbb{R}$ is $\varepsilon$-Hölder continuous: $\forall x, x^{\prime} \in \mathbb{R}$

$$
\left|g(x)-g\left(x^{\prime}\right)\right| \leq\|g\|_{\varepsilon}\left|x-x^{\prime}\right|^{\varepsilon} .
$$

(2) $f:[0, T] \times \mathbb{R} \times \mathbb{R} \times \mathbb{R} \longrightarrow \mathbb{R} \forall(t, x, y, z)$ and $\left(t^{\prime}, x^{\prime}, y^{\prime}, z^{\prime}\right)$

$$
\begin{aligned}
& \left|f(t, x, y, z)-f\left(t^{\prime}, x^{\prime}, y^{\prime}, z^{\prime}\right)\right| \\
& \leq\left\|f_{t}\right\|_{\alpha}\left|t-t^{\prime}\right|^{\alpha}+\left\|f_{x}\right\|_{\varepsilon}\left|x-x^{\prime}\right|^{\varepsilon}+\|f\|_{\text {Lip }}\left(\left|y-y^{\prime}\right|+\left|z-z^{\prime}\right|\right) .
\end{aligned}
$$

## Theorem 3 (Briand, C.G., S.Geiss, Labart (2021))

Assume that Assumption 1 holds. Then there exists a constant $C_{\varepsilon}>0$, depending at most on $(T, \alpha, \varepsilon, f, g)$ such that for all $x \in \mathbb{R}$,
(1) $W_{\psi_{1 / \varepsilon}}\left(Y_{s}^{n, t, x}, Y_{s}^{t, x}\right) \leq C_{\varepsilon}(1+|x|)^{\varepsilon} n^{-\left(\alpha \wedge \frac{\varepsilon}{2}\right)} \quad$ for all $0 \leq t \leq s \leq T$,
(11) $W_{\psi_{1 / \varepsilon}}\left(Z_{s}^{n, t, x}, Z_{s}^{t, x}\right) \leq C_{\varepsilon} \frac{(1+|x|)^{\varepsilon}}{\sqrt{T-s}} n^{-\left(\alpha \wedge \frac{\varepsilon}{2}\right)} \quad$ for all $s \in[t, T[$.

In particular, for any $p \in\left[1, \infty\left[\right.\right.$, there exists a constant $C_{p}>0$, depending at most on ( $T, \alpha, \varepsilon, f, g, p$ ) such that for all $x \in \mathbb{R}$,
(1) $W_{p}\left(Y_{s}^{n, t, x}, Y_{s}^{t, x}\right) \leq C_{p}(1+|x|)^{\varepsilon} n^{-\left(\alpha \wedge \frac{\varepsilon}{2}\right)} \quad$ for all $0 \leq t \leq s \leq T$,
(11) $W_{p}\left(Z_{s}^{n, t, x}, Z_{s}^{t, x}\right) \leq C_{p} \frac{(1+|x|)^{\varepsilon}}{\sqrt{T-s}} n^{-\left(\alpha \wedge \frac{\varepsilon}{2}\right)} \quad$ for all $s \in[t, T[$.

The rate is optimal: If $g(x)=x$ and $f=0$ we have $Y_{T}^{n, 0,0}=B_{T}^{n}$ and $Y_{T}^{0,0}=B_{T}$, and it holds

$$
\lim _{n \rightarrow \infty} n^{\frac{1}{2}} W_{1}\left(Y_{T}^{n, 0,0}, Y_{T}^{0,0}\right) \geq \frac{1}{2} T^{\frac{1}{2}}
$$

Wasserstein convergence rates for $\left(Y^{n}, Z^{n}\right) \rightarrow(Y, Z)$

Idea of the proof: using the connection
FBSDE $\Longleftrightarrow$ semilinear heat equation
$\mathrm{FBS} \Delta \mathrm{E} \quad \Longleftrightarrow \quad$ finite difference equation

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FBSDE $\Longleftrightarrow$ semilinear heat equation convergence $\Uparrow$ 介
$\mathrm{FBS} \Delta \mathrm{E} \quad \Longleftrightarrow \quad$ finite difference equation

## FBSDE $\Longleftrightarrow$ semilinear heat equation

$$
\left\{\begin{array}{l}
\partial_{t} u(t, x)+\frac{1}{2} \Delta u(t, x)+f(t, x, u(t, x), \nabla u(t, x))=0, \\
u(T, \cdot)=g
\end{array} \quad(t, x) \in[0, T) \times \mathbb{R}\right.
$$

is associated to the FBSDE

$$
\begin{gathered}
Y_{t}=g\left(B_{T}\right)+\int_{t}^{T} f\left(B_{s}, Y_{s}, Z_{s}\right) d s-\int_{t}^{T} Z_{s} d B_{s}, \quad 0 \leq t \leq T \\
\Leftarrow: \quad Y_{s}=u\left(s, B_{s}\right) \text { and } Z_{s}=\nabla u\left(s, B_{s}\right) \\
\Rightarrow: \\
u(t, x)=\mathbb{E}\left[g\left(B_{T}^{t, x}\right)+\int_{t}^{T} f\left(r, B_{r}^{t, x}, Y_{r}^{t, x}, Z_{r}^{t, x}\right) d r\right] \\
\nabla u(t, x)=\mathbb{E}\left[g\left(B_{T}^{t, x}\right) \frac{B_{T}-B_{t}}{T-t}+\int_{t}^{T} f\left(r, B_{r}^{t, x}, Y_{r}^{t, x}, Z_{r}^{t, x}\right) \frac{B_{r}-B_{t}}{r-t} d r\right] \\
\quad(t, x) \in[0, T) \times \mathbb{R}
\end{gathered}
$$

For the semilinear heat equation

$$
\left\{\begin{array}{l}
\partial_{t} u(t, x)+\frac{1}{2} \Delta u(t, x)+f(t, x, u(t, x), \nabla u(t, x))=0, \quad(t, x) \in[0, T) \times \mathbb{R} \\
u(T, \cdot)=g,
\end{array}\right.
$$

we define the finite difference scheme letting

$$
\begin{aligned}
L^{h} u(x) & :=\frac{1}{2 h}(u(x+\sqrt{h})+u(x-\sqrt{h})-2 u(x)) \\
\nabla^{h} u(x) & :=\frac{1}{2 \sqrt{h}}(u(x+\sqrt{h})-u(x-\sqrt{h})) \\
\partial_{t}^{h} u(t) & :=\frac{1}{h}(u(t+h)-u(t))
\end{aligned}
$$

and $t_{k}=k h$. Then

$$
\left\{\begin{array}{lc}
\partial_{t}^{h} U^{n}\left(t_{k}, x\right)+L^{h} U^{n}\left(t_{k+1}, x\right)+f\left(t_{k+1}, x, U^{n}\left(t_{k}, x\right), \nabla^{h} U^{n}\left(t_{k+1}, x\right)\right)=0, \\
U^{n}\left(t_{n}, x\right)=g(x) . & k=0, \ldots, n-1
\end{array}\right.
$$

## $\mathrm{FBS} \Delta \mathrm{E} \Longleftrightarrow$ finite difference equation

finite difference equation

$$
\begin{cases}\partial_{t}^{h} U^{n}\left(t_{k}, x\right)+L^{h} U^{n}\left(t_{k+1}, x\right)+f\left(t_{k+1}, x, U^{n}\left(t_{k}, x\right), \nabla^{h} U^{n}\left(t_{k+1}, x\right)\right)=0, \\ U^{n}\left(t_{n}, x\right)=g(x) . & k=0, \ldots, n-1\end{cases}
$$

FBS $\Delta \mathrm{E}$

$$
\begin{gathered}
\left\{\begin{array}{l}
Y_{t_{k}}^{n}=Y_{t_{k+1}}^{n}+h f\left(t_{k+1}, B_{t_{k}}^{n}, Y_{t_{k}}^{n}, Z_{t_{k+1}}^{n}\right)-\sqrt{h} Z_{t_{k+1}}^{n} \varepsilon_{k+1} \\
Y_{T}^{n}=g\left(B_{T}^{n}\right), \quad k=0, \ldots, n-1
\end{array}\right. \\
\Leftarrow: \quad Y_{t_{k}}^{n}=U^{n}\left(t_{k}, B_{t_{k}}^{n}\right), \quad Z_{t_{k}}^{n}=\nabla^{h} U^{n}\left(t_{k}, B_{t_{k-1}}^{n}\right), \\
\Rightarrow: \\
U^{n}(t, x)=\mathbb{E} g\left(B_{T}^{n, t, x}\right)+\mathbb{E} \int_{(t, T]} f\left(s, B_{s-}^{n, t, x}, Y_{s-}^{n, t, x}, Z_{s}^{n, t, x}\right) d\left[B^{n}\right]_{s} \quad 0 \leq t \leq T . \\
\nabla^{h} U^{n}(\underline{t}+h, x)=\mathbb{E}\left[g\left(B_{T}^{n, t, x}\right) \frac{B_{T}^{n, t, x}-x}{T-\underline{t}}+\int_{(t, T]} f(\ldots) \frac{B_{s}^{n, t, x}-x}{\underline{s}-\underline{t}} d\left[B^{n}\right]_{s}\right]
\end{gathered}
$$

## Theorem 4

Under Assumption 1 there exists a constant $C>0$ depending at most on ( $T, \alpha, \varepsilon, f, g$ ) such that for $k=0, \ldots, n-1$
(1) $\left|u(t, x)-U^{n}\left(t_{k}, x\right)\right| \leq C(1+|x|)^{\varepsilon} n^{-\left(\alpha \wedge \frac{\varepsilon}{2}\right)} \quad$ for all $(t, x) \in \mathbb{R} \times\left[t_{k}, t_{k+1}\right)$,
(11) $\left|\nabla u(t, x)-\nabla^{h} U^{n}\left(t_{k+1}, x\right)\right| \leq C \frac{(1+|x|)^{\varepsilon}}{\sqrt{T-t}} n^{-\left(\alpha \wedge \frac{\varepsilon}{2}\right)}$ for all $(t, x) \in \mathbb{R} \times\left[t_{k}, t_{k+1}\right)$.

Open question: the path dependent case (How to construct Malliavin weights?)
First answer: approximation of the Brownian motion by random walk without Skorohod embedding.

Theorem 5 (Briand, C.G., S.Geiss, Labart)
$\left(B_{t}\right)_{t \in[0, T]}$ on $(\Omega, \mathcal{F}, \mathbb{P})$. Then one can construct
(1) an extension $\left(\Omega^{\prime}, \mathcal{F}^{\prime}, \mathbb{P}^{\prime}\right)$
(2) $\varepsilon_{1}, \ldots, \varepsilon_{n}$ independent Rademacher $r v$ such that for $p \in(0, \infty)$

$$
\left(\mathbb{E}_{\mathbb{P}^{\prime}} \sup _{t \in[0, T]}\left|B_{t}-B_{t}^{n}\right|^{p}\right)^{1 / p} \leq C \frac{\log n}{\sqrt{n}}
$$

P. Briand, B. Delyon, J. Memin, Donsker-Type theorem for BSDEs. Electron. Comm. Probab., 6 1-14 (2001).
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P. Briand, C. Geiss, S. Geiss, C. Labart, Donsker-Type Theorem for BSDEs: Rate of Convergence. Bernoulli, 27(2): 899-929 (2021).
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