ERRATUM

## **Erratum to: Rough Burgers-like equations** with multiplicative noise

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## Erratum to: Probab. Theory Relat. Fields (2013) 155:71–126 DOI 10.1007/s00440-011-0392-1

Unfortunately, the proof of Proposition 4.8 in the original article is not correct. In fact, equation (4.60), where  $R^{\theta}$  is rewritten as a stochastic integral, has no meaning because the integrand is not adapted. Hence in (4.61) we are not allowed to apply the Burkholder–Davies–Gundy inequality.

The argument was corrected in [1, Lemma 3.6 and Corollary 3.7] in a more complicated situation. In order to transfer these statement to the situation in the original article some small changes are needed. First of all, the regularity of the linearised process *X* should be measured in a Hölder norm with slightly bigger index  $\alpha_{\star}$  than the solution *u*. This can be done without further problems. With this change, the definition of the stopping time  $\tau_K^X$  in (4.53) should be replaced by the following.

*For* K > 0 *and for an*  $\alpha_{\star} \in (\alpha, 1/2)$  *we introduce the stopping time* 

$$\tau_K^X = \inf\left\{ t \in [0, T] \colon \sup_{\substack{x_1 \neq x_2 \\ 0 \leq s_1 < s_2 \leq t}} \frac{\left| X(s_1, x_1) - X(s_2, x_2) \right|}{|s_1 - s_2|^{\alpha_\star/2} + |x_1 - x_2|^{\alpha_\star}} > K \right\}.$$
 (1)

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With this changed definition, Lemma 3.6 of [1] implies, in the notation of the original article, the following result.

**Lemma 1** Suppose that  $0 < \alpha < \alpha_{\star} < \frac{1}{2}$  and let  $\tau$  be a stopping time that almost surely satisfies

$$0 \leq \tau \leq \tau_K^X \wedge T.$$

For every  $0 \le t \le T$  we set

$$\begin{split} \tilde{\theta}(t) &:= \theta(t \wedge \tau), \\ \tilde{\Psi}^{\theta}(t) &:= \int_{0}^{t \wedge \tau} S(t-r) \,\tilde{\theta}(r) \, dW(r), \\ \tilde{X}(t) &:= \int_{0}^{t \wedge \tau} S(t-r) \, dW(r), \\ \tilde{R}^{\theta}(t; x, y) &:= \delta \tilde{\Psi}^{\theta}(t; x, y) - \tilde{\theta}(t, x) \, \delta \tilde{X}(t; x, y) \end{split}$$

Then, for any p large enough and for any  $\gamma > 0$  such that

$$\gamma < \alpha_{\star} + \alpha - \frac{1}{p} - \sqrt{\frac{1}{2p}(1 + \alpha - \alpha_{\star})},$$

the following bound holds true:

$$\sup_{0 < t \le T} \mathbb{E} \left| \tilde{R}^{\theta}(t) \right|_{\Omega \mathcal{C}^{\gamma}}^{p} \lesssim \left\| \theta \right\|_{p,\alpha}^{p}.$$
<sup>(2)</sup>

The statement given here is actually slightly stronger than the bound stated in [1] because the norm appearing on the left hand side of (2) is bounded uniformly in t instead of allowing a blow up near 0. In [1] we had to introduce this blowup due to a slightly modified definition of the Gaussian process X: the process used in [1] does not start at 0, but with stationary initial condition, which was convenient for other reasons. When going through the proof given in [1], one realises that when considering the process with zero initial condition, one can apply bound (3.74) for all times t and there is no need to use (3.75) for small times.

Based on this version of Lemma 1, it is then straightforward to use the a priori information on the time regularity of  $R^{\theta}$ , combined with the fact that the "tilde" processes coincide with the "non-tilde" processes before time  $\tau$ , to obtain the bound

$$\mathbb{E}\left[\left\|R^{\theta}\right\|_{C^{\kappa}\left([0,\tau];\Omega C^{2\alpha}\right)}^{p}\right] \leq C(K,T)\left\|\theta\right\|_{p,\alpha}^{p},$$

for sufficiently small values of  $\kappa$  and sufficiently large values of p, as required.

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## Reference

1. Hairer, M., Maas, J., Weber, H.: Approximating rough stochastic PDEs (arXiv e-prints)