An introduction to singular SPDEs

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Dedicated to Michael Röckner on the occasion of his 60th Birthday.

Abstract

We review recent results on the analysis of singular stochastic partial differential equations in the language of paracontrolled distributions.

1 Introduction

In recent years there has been much progress in the mathematical understanding of certain non-linear random PDEs which are not well posed in the classical analytic or probabilistic theory but which become amenable to rigorous analysis as soon as specific non-linear properties of the randomness are taken into account. In the related literature it has become common to refer to such equations as singular SPDEs, mainly to distinguish them from standard SPDEs. The difference is that singular SPDEs can be posed only in small subspaces of the standard function spaces (e.g. Hölder, Sobolev or even Besov spaces) and that the operations involved in such equations sometimes require renormalisation. Renormalisation in this context can be understood as the fact that only specific non-linearities can be formed meaningfully and that, in order to do so, subtractions of infinite quantities are often needed.

The aim of this short review is to present some of the ideas underlying singular SPDEs and their pathwise analysis and to make the connection with other parts of probability theory and mathematical physics: to renormalisation group theory and to scaling limits of interacting particle models. An important motivation to consider singular SPDEs is indeed that in some cases they appear in the description of universal large scale fluctuations for certain spatially extended probabilistic models. Universality here means that irrespective of most specific features of the model its large scale fluctuations are described by a generic equation that usually can be guessed by first principles and then hopefully confirmed by rigorous analysis. One of the main open problem in singular SPDEs is to enlarge the spectrum of models for which the universality can be rigorously proven.

Currently we dispose of four main approaches in order to study singular equations: regularity structures, paracontrolled distributions, the renormalisation group approach, and Otto’s and Weber’s rough path based approach. Apart from the renormalisation

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group approach, the other three techniques are all inspired by T. Lyons’s rough path theory \cite{Lyo98, LQ02, LCL07} and by the related notion of controlled paths \cite{Gub04, FH14}. Regularity structures have been introduced by M. Hairer in his remarkable work \cite{Hai14} where they were in particular used to give, for the first time, a solution theory for the dynamic $\Phi^4_3$ model. Regularity structures allow a detailed description of the local action of distributions on test functions in terms of a given model which usually is constructed from certain non-linear features of an underlying random process. Paracontrolled distributions have been introduced by the authors together with P. Imkeller \cite{GIP15}, more or less at the same time as M. Hairer was developing his theory, as a tool to describe the “spectral” features of a function (or distribution) in terms of simpler objects, very similar to Hairer’s models. Some time after, Kupiainen \cite{Kup16} observed that the renormalisation group features of a function (or distribution) in terms of simpler objects, very similar to Hairer’s.

We will illustrate the analysis of singular SPDEs on a series of models:

1. The 1d generalised Stochastic Burgers equation (gSBE)
\[
\partial_t u(t, x) = \Delta u(t, x) + G(u(t, x))\partial_x u(t, x) + \xi(t, x), \quad t \geq 0, x \in \mathbb{T},
\]
where $u: \mathbb{R}_+ \times \mathbb{T} \to \mathbb{R}$ and $G: \mathbb{R} \to \mathbb{R}$ is a smooth function.

2. The 1d Kardar–Parisi–Zhang equation (KPZ)
\[
\partial_t h(t, x) = \Delta h(t, x) + (\partial_x h(t, x))^2 - C + \xi(t, x), \quad t \geq 0, x \in \mathbb{T}
\]
with $h: \mathbb{R}_+ \times \mathbb{T} \to \mathbb{R}$ and $C \in \mathbb{R}$ and the related conservative stochastic Burgers equation (CSBE)
\[
\partial_t u(t, x) = \Delta u(t, x) + \partial_x (u(t, x)^2) + \partial_x \xi(t, x), \quad t \geq 0, x \in \mathbb{T},
\]
where $u = \partial_x h: \mathbb{R}_+ \times \mathbb{T} \to \mathbb{R}$.

3. The dynamic $\Phi^3_4$ model or stochastic quantisation equation ($d = 2, 3$) (SQE)
\[
\partial_t \varphi(t, x) = \Delta \varphi(t, x) - \varphi^3(t, x) + C\varphi(t, x) + \xi(t, x), \quad t \geq 0, x \in \mathbb{T}^d,
\]
where $\varphi: \mathbb{R}_+ \times \mathbb{T}^d \to \mathbb{R}$ and $C \in \mathbb{R}$.

4. The generalised two-dimensional parabolic Anderson model (gPAM)
\[
\partial_t u(t, x) = \Delta u(t, x) + G(u(t, x))\xi(x) - CG'(u(t, x))G(u(t, x)), \quad t \geq 0, x \in \mathbb{T}^2,
\]
where $u: \mathbb{R}_+ \times \mathbb{T}^2 \to \mathbb{R}$, $G: \mathbb{R} \to \mathbb{R}$ is a smooth function, and $C \in \mathbb{R}$.

In all these examples $\mathbb{T}^d = \mathbb{R}^d / \mathbb{Z}^d$ is the $d$-dimensional torus and $\xi$ denotes a Gaussian white noise (space–time or space–dependent only, according to the model). The specific choice of the dimensionality and/or space–time dependence of the noise is related to the degree of singularity of the equation and is motivated by the fact that in this review we will use the specific language of paracontrolled distributions which has the advantage of requiring very little background and is more directly related to standard PDE theory than the other approaches.

This choice has the drawback that we will not address the discussion of natural models like the generalised form of the KPZ equation, given by
\[
\partial_t h(t, x) = \Delta h(t, x) + G(h(t, x))(\partial_x h(t, x))^2 + F(h(t, x))\xi(t, x), \quad t \geq 0, x \in \mathbb{T},
\]
which is within reach of regularity structure theory but still out of reach for paracontrolled distributions. The generalised KPZ equation is in a way the “ultimate” singular SPDE and its singularities are very challenging. Its analysis via regularity structures requires a great deal of work and a deeper understanding of the algebraic and analytic structures underlying the construction of an appropriate model and of the renormalisation [BHZ16, CH16].

As we already remarked the main difficulty shared by all singular SPDEs is the presence of non-linear operations which are not well defined in classical function spaces. This difficulty blocks the analysis from the very beginning because it is not even possible to formulate the equation rigorously. All the equations we wrote in the introduction are classically ill-posed and the notation is just used in an informal and suggestive way. Indeed most of them have to be modified to take into account renormalisation which means that the constant $C$ that appear in the equations is actually to be read as an infinite quantity and not a real number. A standard approach to make the analysis rigorous is to construct a series of approximate problems whose solutions converge to a well-defined limit. The characterisation of this limit and its independence of the details of the approximation procedure will constitute a rigorous definition of the solution to a singular SPDE. Taking as an example the 2d gPAM, we can (loosely) state a typical convergence result:

**Theorem 1.1** Let $G : \mathbb{R} \rightarrow \mathbb{R}$ be a smooth function, $\varepsilon > 0$, and let $u_\varepsilon : \mathbb{R}^+ \times \mathbb{T}^2 \rightarrow \mathbb{R}$ be the unique classical solution of the Cauchy problem

$$
\partial_t u_\varepsilon(t, x) = \Delta u_\varepsilon(t, x) + G(u_\varepsilon(t, x))\xi_\varepsilon(x) - G_2(u_\varepsilon(t, x))c_\varepsilon,
$$

where $\xi_\varepsilon$ is a smooth approximation of the space white noise $\xi : \mathbb{T}^2 \rightarrow \mathbb{R}$ by convolution, $c_\varepsilon \in \mathbb{R}$, and $G_2(u) = G'(u)G(u)$. Then there exists a choice for $c_\varepsilon$ such that locally in time the sequence $u_\varepsilon$ converges, as $\varepsilon \to 0$, to a function $u$ that does not depend on the exact form of the convolutional approximation of $\xi_\varepsilon$.

This formulation is quite minimal, in reality the analysis which has to be put forward to prove this kind of results give as a byproduct also much more detailed information about the function $u$. In particular it is possible to characterise $u$ via a standard PDE for a different unknown and the assumptions on the approximations of the noise $\xi$ become assumptions about the convergence of certain non-linear functionals of $\xi_\varepsilon$. The general scheme underlying the convergence is the following:

$$
\begin{array}{cccc}
\xi_\varepsilon & \xleftarrow{J} & \Xi_\varepsilon & \xrightarrow{\Phi} & U_\varepsilon & \xrightarrow{\Pi} & u_\varepsilon \\
\downarrow & & \downarrow & & \downarrow & & \downarrow \\
\xi & \leftarrow \Xi & \xrightarrow{\Phi} & U & \xrightarrow{\Pi} & u
\end{array}
$$

Here the vertical arrows represent limits for $\varepsilon \to 0$ (in appropriate function spaces). The upper row features $\Xi_\varepsilon = J(\xi_\varepsilon)$, an injective collection of non-linear quantities constructed from the approximate data $\xi_\varepsilon$ (one could also consider the initial condition to be part of this data but we will refrain from doing so), and $U_\varepsilon = \Phi(\Xi_\varepsilon)$, an enhanced notion of solution from which one can recover the classical solution $u_\varepsilon$ through a continuous projection $\Pi$. The bottom row describes the situation after the limit $\varepsilon \to 0$ has been taken. The enhanced data $\Xi$ still determines the limit noise $\xi$, however the reverse in not true and different approximation procedures for the same $\xi$ can lead to different values of $\Xi$. But the remaining relations are preserved: from $\Xi$ we can still recover an enhanced notion of solution $U$ through the same continuous solution map $\Phi : \Xi \rightarrow U$, and from $U$ we obtain $u$ by a projection in such a way that the convergence $u_\varepsilon \rightarrow u$ holds by the continuity of the solution map and of the projection. In the limit the situation is more complex than before passing to the limit. As mentioned before a different approximation
procedure $\tilde{\xi}_\varepsilon \to \xi$ can lead to different enhanced data $\tilde{\Xi}$ in the limit, and as a consequence of the continuity of $\Phi$ to a different limiting solution $\tilde{u}$. This situation is depicted in the graph below.

So the relation between the data $\xi$ and the solution is not well defined unless some information on the non-linear features $\Xi$ is provided as additional input.

Proving a convergence result therefore requires two conceptually different steps:

1. define and analyse the continuity of the map $\Phi$, where an important task is to identify its domain and co-domain;
2. prove the convergence of the enhanced data $\Xi_\varepsilon \to \Xi$.

The convergence of the enhanced data is, in most of the applications, a purely probabilistic step which sometimes requires development of efficient tools but for which the main tools are classical and already present in the probabilistic literature for a long time. Some keywords in this context are Gaussian analysis, hypercontractivity, Besov embeddings, almost sure regularity of stochastic processes, martingale theory, Wick products, and chaos expansions. We would like to concentrate our discussion to the analytic part of the theory involving the construction of the enhanced spaces which constitute the domain and co-domain of the solution map $\Phi$.

In Sections 2 and 3 we present the basic ideas and analytic ingredients for paracontrolled distributions. Section 4 briefly discusses the need for renormalisation of the enhanced data and how this renormalisation translates in the equation. In Section 5 we sketch recent work of Bailleul and Bernicot on higher order expansions via paraproducts. Section 6 is dedicated to convergence results for singular SPDEs and we illustrate how to derive the Hairer-Quastel weak universality principle for the KPZ equation using paracontrolled distributions. In Sections 7, 8 we will see that paracontrolled distributions can not only be used to study singular SPDEs, but as noted by Cannizzaro and Chouk respectively Allez and Chouk they also allow us to construct certain singular operators.

Finally let us point out that there are many fascinating recent results that are based on paracontrolled distributions and that we have to omit here due to space constraints. To name just a few: non-explosion results for the dynamic $\Phi^3_4$ model [MW16], the KPZ equation [GP17], and the multi-component KPZ equation [FH16], a formulation of paracontrolled distributions that allows to study equations on manifolds [BB16a], convergence results for discrete dynamics [ZZ14, ZZ15, GP17, CGP16], a solution theory for quasilinear equations [BDH16], nonlinear extensions of paraproducts [FG16], and a support theorem for gPAM [CF14] – not to mention all the exciting results that have been shown in the setting of regularity structures or the other approaches.

2 Paraproducts

In order to develop the paracontrolled analysis of the solution map we will introduce in this section Besov spaces and paraproducts. See [BCD11] for details. Let $S'(\mathbb{T}^d)$ denote the Schwartz space of distributions on $\mathbb{T}^d = \mathbb{R}^d/\mathbb{Z}^d$. Then any element $f \in S'(\mathbb{T}^d)$ can be
decomposed as

\[ f = \sum_{i \geq -1} \Delta_i f, \]

where the sum is over \( i = -1, 0, 1, 2, \ldots \) and \( \Delta_i f \) are smooth functions whose Fourier support is contained in a ball \( B \subseteq \mathbb{R}^d \) for \( i = -1 \) and in rescaled dyadic annuli \( 2^i A \) for \( i \geq 0 \). The operators \( \Delta_i : f \mapsto \Delta_i f \) are called Littlewood–Paley operators and can be constructed to enjoy nice analytic properties, for example they satisfy the Bernstein inequalities

\[ \|D^\alpha \Delta_i f\|_{L^p(\mathbb{T}^d)} \lesssim 2^{i|\alpha|+\frac{d}{2}} \|\Delta_i f\|_{L^q(\mathbb{T}^d)}, \quad i \geq -1, \quad p \geq q, \]

where \( \alpha \) is a \( d \)-dimensional multi-index and \( D^\alpha \) denotes the related mixed derivative of order \( |\alpha| \). The (inhomogeneous) Besov space \( B^\alpha_{p,q} \) is defined as the set of all distributions \( f \in \mathcal{S}'(\mathbb{T}^d) \) such that the norm

\[ \|f\|_{B^\alpha_{p,q}} = \|(2^\alpha \Delta_i f)\|_{L^p(\mathbb{T}^d)} \]

is finite. In the analysis of singular SPDEs we will mainly use the Hölder–Besov spaces \( C^\alpha = B^\alpha_{\infty, \infty} \) to get rid of the integrability exponents. This turns out to be very convenient for non-linear estimates because \( \|fg\|_{L^\infty} \lesssim \|f\|_{L^\infty} \|g\|_{L^\infty} \) and similarly for \( L^\infty \) which is not true for the \( L^p \) respectively \( L^\infty \) norms with \( p < \infty \). To gain an intuitive understanding of the \( C^\alpha \) spaces it is useful to note that for \( \alpha \in \mathbb{R}_+ \setminus \mathbb{Z} \) the space \( C^\alpha \) consists exactly of the \( \lfloor \alpha \rfloor \) times continuously differentiable functions from \( \mathbb{T}^d \) to \( \mathbb{R} \) for which the classical increment-based \( (\alpha - \lfloor \alpha \rfloor) \)-Hölder norm of all partial derivatives of order \( |\alpha| \) is finite. And roughly speaking for \( \alpha \in (-1,0) \) every \( f \in C^\alpha \) is the distributional derivative of some \( F \in C^{\alpha+1} \), and similarly distributions of lower regularity are higher order derivatives of Hölder continuous functions.

The Littlewood–Paley decomposition induces a natural decomposition of products of Besov functions in terms of paraproducts. Given \( f, g \) we have

\[ fg = \sum_{i,j \geq -1} \Delta_i f \Delta_j g = f \ll g + f \gg g + f \bowtie g, \]

where the paraproducts \( f \ll g \) and \( f \gg g \) and the resonant product \( f \bowtie g \) by are defined, respectively, by

\[ f \ll g = g \gg f := \sum_{i \geq -1} (\Delta_{\ll i-1} f) \Delta_{i} g, \quad f \gg g := \sum_{i,j : |i-j| \leq 1} \Delta_i f \Delta_j g, \]

and where we introduced the notation \( \Delta_{<k} f = \sum_{\ell < k} \Delta_\ell f \). Paraproducts are continuous bilinear operations on the following function spaces

\[ \star \ll \star : C^\alpha \times C^\beta \to C^{\beta+\alpha}, \quad \alpha \leq 0, \beta \in \mathbb{R}, \]
\[ \star \gg \star : L^\infty \times C^\beta \to C^\beta, \quad \beta \in \mathbb{R}, \]

while the resonant product is well defined only if \( \alpha + \beta > 0 \) and in that case it is a continuous bilinear operator

\[ \star \bowtie \star : C^\alpha \times C^\beta \to C^{\beta+\alpha}. \]

In particular we see that the usual product can be extended by continuity from smooth functions to Hölder–Besov distributions as a bilinear map

\[ (f,g) \mapsto f g : C^\alpha \times C^\beta \to C^{\min(\alpha,\beta)} \]

provided \( \alpha + \beta > 0 \).
Singular SPDEs are characterised by this condition not being satisfied in the non-linear terms. The products then become problematic and undefined for general inputs. To explain the difficulty let us note that given \( \alpha, \beta \in \mathbb{R} \) with \( \alpha + \beta < 0 \) and given smooth functions \((f, g)\) it is easy to construct sequences of functions \((f_n, g_n)\) such that \((f_n, g_n) \to (f, g)\) in \( C^\alpha \times C^\beta \) and such that the pointwise limit \( \lim_n f_n g_n \) exists and is smooth but nonetheless
\[
\lim_n f_n g_n \neq fg.
\]
If we want a “robust” way to consider the product \( fg \) in the situation \( \alpha + \beta < 0 \), then we should take this ambiguity into account from the start and think about the product as describing a manifold of possibilities and not just a single deterministic operation on the inputs.

Part of the analysis of singular SPDEs can be understood as a classification of these ambiguities: we track the extent to which the possible outcomes of undefined operations can propagate into the solution theory of a given equation. We will come back to this point below when we discuss renormalisations.

Paraproducts and related paradifferential operators have been introduced in the seminal work of Bony on the propagation of singularities for non-linear hyperbolic equations [Bon81, Mey81]. They provide a good approximation of non-linear operations, in this case the product, but can be used also to linearise other operations. For example, the following paralinearisation result is useful in order to deal with equation with non-polynomial coefficients: Given a smooth function \( f : \mathbb{R} \to \mathbb{R} \) we have
\[
u \in C^\alpha \mapsto R_f(u) := f(u) - f'(u) u \in C^{2\alpha}, \quad \alpha > 0,
\]
which shows that the composition \( f(u) \in C^\alpha \) behaves like the paraproduct \((f'(u) u) \in C^\alpha \) modulo a smoother correction term.

One main tool in the paracontrolled analysis is the following commutator lemma which describes the interaction between the resonant product and the paraproduct. For a proof see [GIP15].

**Lemma 2.1** Let \( \alpha \in (0, 1), \beta, \gamma \in \mathbb{R} \) be exponents such that \( \beta + \gamma < 0 \) and \( \alpha + \beta + \gamma > 0 \). Then there exists a continuous trilinear map \( C : C^\alpha \times C^\beta \times C^\gamma \to C^{\alpha + \beta + \gamma} \) such that if \( f, g, h \) are smooth functions we have
\[
C(f, g, h) = (f \prec g) h - f(g \circ h).
\]

### 3 Paracontrolled analysis

The construction of the solution maps proceeds via perturbative analysis with respect to a linearised approximation of the equation. We look for an expansion in terms of regularities of the various terms. In the following we ignore the initial conditions for the equations that we treat and when considering a solution \( w \) to \( \mathcal{L}w = v \) we always silently assume that \( w(0) = 0 \). Once this is understood the case of general initial conditions does not add much conceptual (although some technical) difficulty. We also ignore the need for renormalisation at the moment and set the (infinite) constants \( C \) appearing in the equations equal to 0 in the following discussion. In general we omit many technical details, an introduction to paracontrolled distributions that provides more details can be found in the lecture notes [GP15].

Among the examples we treat, the simplest is the dynamic \( \Phi^4_2 \) model (4), \( \varphi : \mathbb{R}_+ \times \mathbb{T}^2 \to \mathbb{R}, \)
\[
\partial_t \varphi(t, x) = \Delta \varphi(t, x) - \varphi(t, x)^3 + \xi(t, x), \quad t \geq 0, x \in \mathbb{T}^2.
\]
This equation was first solved by Albeverio and Röckner [AR91] in 1991, a solution that is more related to the tools we present here was given in 2003 by Da Prato and Debussche [DPD03], and the link between the two solution concepts was recently understood by Röckner, Zhu, and Zhu [RZZ15]. Da Prato and Debussche observed that we can decompose the solution as a sum of two terms which we postulate to be of increasing regularities: Setting \( \varphi = X + \psi \), the equation takes the form

\[ LX + L\psi = -X^3 - 3X^2\psi - 3X\psi^2 - \psi^3 + \xi \]

where we use the notation \( L = \partial_t - \Delta \) and we choose \( X \) in order to cancel the most irregular term on the right hand side, namely the additive noise. So if \( LX = \xi \), a simple inhomogeneous linear equation that can be explicitly solved by convolving \( \xi \) against the heat kernel, we are left with an equation for \( \psi \),

\[ L\psi = -X^3 - 3X^2\psi - 3X\psi^2 - \psi^3. \tag{7} \]

Using stochastic analysis it is possible to prove that

\[ X \in C_T^{\alpha_-} := \bigcap_{\delta > 0} C([0, T], C^{\delta}), \]

and therefore the right hand side of the equation for \( \psi \) is still not well defined because it features the products \( X^2 \) and \( X^3 \) of the distribution \( X \) of negative regularity. For the moment we assume that \( X^3, X^2 \) exist as elements of \( C_T^{\alpha_-} \). They will be part of the non-linear features needed to define the solution map. If we also assume that

\[ \psi \in C_T^{\alpha_+} := \bigcup_{\delta > 0} C([0, T], C^{\delta}), \]

then the right hand side of eq. (7) is well defined and the estimates for the product give us \( X^2\psi, X\psi^2 \in C_T^{\alpha_-} \). By standard estimates for the heat semigroup generated by the Laplace operator \( \Delta \) the map that sends \( v \) to the solution \( w \) of \( \mathcal{L}w = v \) is continuous from \( C_T^{\alpha} = C([0, T], C^{\alpha}) \) equipped with \( \|v\|_{C_T^{\alpha}} = \sup_{t \in [0, T]} \|v(t)\|_{C^\alpha} \) to \( C_T^{\alpha+2} \). So it follows that we can pose the equation (7) for \( \psi \) as a standard PDE with unique weak solution \( \psi \in C_T^{C^{2-}} \) for some\(^1 \) \( T > 0 \). Here the change of variable has therefore been quite simple: The enhancement \( J \), the solution map \( \Phi \) and the projection \( \Pi \) take the form

\[ J : \xi \mapsto (X, X^2, X^3), \]

\[ \Phi : \Xi = (\zeta_1, \zeta_2, \zeta_3) \mapsto U = (\zeta_1, \psi), \quad \Pi : U = (\zeta_1, \psi) \mapsto \varphi = \zeta_1 + \psi, \]

where \( \psi \) is a weak solution to

\[ \mathcal{L}\psi = \zeta_3 + 3\zeta_2\psi + 3\zeta_1\psi^2 + \psi^3. \tag{8} \]

Note that \( \xi \) can be recovered from \( (\zeta_1, \zeta_2, \zeta_3) = J(\xi) \) via \( \xi = LX \). Observe also that if \( \xi \) is a smooth function, the composition \( \Pi \circ \Phi \circ J : \xi \mapsto \varphi \) gives back the classical weak solution to eq. (4). We will come back again later to this equation to discuss the construction of the enhanced data \( \Xi \) which will require a renormalisation in the case of white noise. For details on the analysis that we sketched above see [DPD03].

While the method we just discussed is simple and elegant, the other singular equations that we mentioned in the introduction apart from the dynamic \( \Phi^2 \) model cannot be handled by a simple additive change of variables. Consider for example the generalised

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\(^1\)Possibly \( T \) is quite small because when setting up the Picard iteration we pick up a superlinear estimate so at this point we cannot exclude the possibility that the solution blows up in finite time. To prove existence for all times we have to make use of the sign of the nonlinearity \( -\varphi^3 \) in (4), see [MW15].
Lemma 2.1 which gives us

\[ \mathcal{L}v = G(X + v)\partial_x X + G(X + v)\partial_x v. \]  

(9)

The analysis of the regularity of \( X \) now gives \( X \in C_T C^{1/2-} \) (it is better behaved than before because we are in one space-dimension and the white noise becomes more and more irregular in higher dimensions) and therefore the best we can hope for the right hand side of (9) is that it takes values in \( C_T C^{-1/2-} \) (the regularity of \( \partial_x X \)) which would put \( v \) in \( C_T C^{3/2-} \), that is \( v \) has two degrees of regularity more than the right hand side which follows from the regularising effect of the inversion of \( \mathcal{L} \) that we discussed above. This would in turn mean that \( G(X + v) \in C_T C^{1/2-} \) provided \( G \) is at least \( \mathcal{C}^1 \). In this setting the product \( G(X + v)\partial_x v \) is well defined since \( \partial_x v \) has regularity \( C_T C^{1/2-} \), but \( G(X + v)\partial_x X \) is not since \( \partial_x X \) has negative regularity \( C_T C^{-1/2-} \) and we barely fail to ensure that the sum of the regularities is positive. But no other simple additive subtraction is available and therefore we need to understand better the structure of the problematic product in order to determine sufficient conditions to control it. The paraproduct decomposition gives

\[ G(X + v)\partial_x X = C^{1/2-}((X + v)\partial_x X) + C^{0-}(G(X + v)\partial_x X + G(X + v)\partial_x v), \]

where the underbraces denote the respective regularities of the two paraproducts and the difficulty is isolated in the resonant term. The paralinearisation result (6) applied to \( G(X + v) \) shows that

\[ G(X + v) = G'(X + v)\prec(X + v) + R_G(X + v), \]

and we can decompose the paraproduct on the right hand side into

\[ G'(X + v)\prec(X + v) = G'(X + v)\prec X + G'(X + v)\prec v, \]

which shows that the irregularity of \( G(X + v) \) comes from the paraproduct \( G'(X + v)\prec X \) and we can further isolate the difficulty in the resonant product:

\[ G(X + v)\circ\partial_x X = C^{1-}((G'(X + v)\prec X)\circ\partial_x X + G'(X + v)\prec v + R_G(X + v)\circ\partial_x X), \]

where the last two terms on the right hand side can be controlled by the estimates for the resonant product because here the sum of the regularities is strictly positive. To deal with the remaining ill-defined resonant product we apply the commutator estimate from Lemma 2.1 which gives us

\[ (G'(X + v)\prec X)\circ\partial_x X = C^{0-}(G'(X + v), X, \partial_x X) + G'(X + v)(A_0 \partial_x X), \]

where we continue to denote with the underbrace “!!” the term which is still problematic according to the deterministic a priori regularities. The reader can convince herself that all the other terms are indeed well defined. This analysis allowed us to isolate the singular nature of the product into some non-linear feature pertaining only to the data \( X \). Much like the simpler algebraic analysis allowed by the \( \Phi_4^1 \) model. Similarly we will now assume a prescribed natural regularity for \( X \circ\partial_x X \), namely \( X \circ\partial_x X \in C_T C^{0-} \), and
include this term in the enhanced data for this problem. Then the remaining product
$G'(X + v)(X \circ \partial_x X)$ is well defined because $G'(X + v) \in C_T \mathcal{C}^{1/2-}$ and therefore the sum
of the regularities of the factors is strictly positive. From here we can continue to solve
the equation for $v$ by a Picard iteration. The enhancement $J$, the solution map $\Phi$ and
the projection $\Pi$ now take the form

$$ J : \xi \mapsto (X, X \circ \partial_x X) $$

$$ \Phi : \Xi = (\xi_1, \xi_2) \mapsto U = (\xi_1, v), \quad \Pi : U = (\xi_1, v) \mapsto u = \xi_1 + v, $$

where $v \in C_T \mathcal{C}^{3/2-}$ is the solution to the PDE

$$ \mathcal{L}v = G(\xi_1 + v) \circ \partial_x \xi_1 + G'(\xi_1 + v) \xi_2 + F(\xi_1, v) $$

and where $F(\xi_1, v)$ is a suitable continuous function taking values in $C_T \mathcal{C}^{0-}$. For details
on this equation we refer to [Hai11] and [GIP15].

A further level of complexity is exemplified by the gPAM (5). In this case it is not even
possible to start the analysis by an additive change of variables. The two dimensional space
white noise has regularity $\mathcal{C}^{-1-}$, so the best regularity we can hope for $v$ is $v \in C_T \mathcal{C}^{-1}$
and then the non-linear term $G(u) \xi$ is not well defined. The paraproduct decomposition gives

$$ G(u) \xi = \underbrace{G(u) \langle u \rangle \xi + G(u) \circ \xi + G(u) \rangle \xi}_{C_T \mathcal{C}^{0-}} $$

and proceeding by paralinearisation and commutation we obtain the following decomposition
of the resonant term

$$ G(u) \langle u \rangle \xi = \underbrace{G'(u) \langle u \rangle \circ \xi + R_G(u) \circ \xi}_{C_T \mathcal{C}^{-1-}} + \underbrace{G(u) \langle u \rangle \circ \xi + R_G(u) \circ \xi}_{C_T \mathcal{C}^{0-}} $$

where we note that $u \circ \xi$ is still not well defined but if we assume it has its natural regularity
$u \circ \xi \in C_T \mathcal{C}^{0-}$, then the product $G'(u) \langle u \rangle \circ \xi$ poses no problem. This means that we can
control the product $G(u) \xi$ once we have a control of the resonant term $u \circ \xi$. Contrary
to the simpler analysis of the gSBE this term is still quite complex since involves the
unknown $u$ and cannot be “postulated” as we did with $X \circ \partial_x X$ before. However, our
analysis shows that the right hand side of the equation can be decomposed in a series
of terms of different regularities, where the worst is $G(u) \langle \xi \rangle$ (assuming for $u \circ \xi$ a better
regularity). So the solution should satisfy

$$ \mathcal{L}u = G(u) \langle \xi \rangle + \cdots, $$

where we neglected more regular terms. The idea is then to make a change of variables
to remove this irregular term in the right hand side. A natural approach is to look for
$u$ with a similar form as the right hand side of the equation, namely a paraproduct plus
a smoother remainder, $u = v \langle X + v \rangle$, where $v \in C_T \mathcal{C}^{1-}$, $X \in C_T \mathcal{C}^{1-}$, $v^2 \in C_T \mathcal{C}^{2-}$ are functions to be determined with $v$ more regular than $X$ and $u$. In order to perform this
change of variables in the equation we need to modify the paraproduct $\ast \langle \ast \rangle$ a bit by
introducing some time-smoothing and defining a modified paraproduct $\ast \ast \ast$ in terms of
which we make the paraccontrolled Ansatz:

$$ u = v \langle X + v \rangle. $$

This modification of the paraproduct is a small technical point which is not very relevant
to the overall picture. The essential property of $\ast \ast \ast$ is that the operator

$$ (f, g) \mapsto H(f, g) = \mathcal{L}(f \ast \ast g) - f \ast \mathcal{L}g $$

\[\text{9}\]
maps\(^2\) the space \(C_T^1 \times C_T^1\) to \(C_T^0\) despite the fact that both summands only live in \(C_T^{-1}\). For the usual paraproduct \(*\prec*\) this is not possible because if we expand \(L(f \prec g) - f \prec Lg\) using Leibniz’s rule we pick up the term \(\partial f \prec g\) which cannot be controlled in terms of the \(C_T^1\) norm of \(f\). The modified paraproduct overcomes this difficulty, but there exist other solutions: either use space-time parabolic Besov spaces and related paraproducts (for which the commutator of paraproduct and \(L\) can be controlled by standard estimates), or define a paraproduct which intertwines exactly with the heat kernel so \(H(f, g) = 0\) (we will discuss this last strategy in more detail in Section 5). In the following we will mostly neglect the difference between these two paraproducts and always consider \(H(f, g)\) as a smoother remainder term. With this proviso we can expand both sides of the equation and get
\[
Lu = v \prec LX + H(v, X) + L\dot{v}^3 = G(u) \prec \xi + \tilde{F}(u, \xi),
\]
where \(\tilde{F}(u, \xi)\) denotes terms that should be in \(C_T^0\). Choosing \(v = G(u)\) and \(LX = \xi\) we can get rid of the irregular term \(G(u) \prec \xi\) on the right hand side and obtain an equation for \(v^3\) which sets it in the good space \(C_T^2\). Now that we have a better description of the solution \(u\) via the paraprocontrolled Ansatz we can go back to the analysis of the resonant term \(u \circ \xi\) and observe that the commutator lemma gives
\[
\begin{align*}
\varphi = (G(u) \circ X) \circ \xi &+ v^2 \circ \xi = G(u) (X \circ \xi) + C(G(u), X, \xi) + v^2 \circ \xi \\
\varphi &
\end{align*}
\]
which again reduces the well-posedness of the right hand side to that of a polynomial non-linear feature constructed from \(\xi\), in this case \(X \circ \xi\). We will assume that \(X \circ \xi\) is given as an element of \(C_T^0\) so in particular the product \(G(u)(X \circ \xi)\) is then well defined and in \(C_T^0\). Resuming this analysis we can conclude that the enhancement \(J\), the solution map \(\Phi\) and the projection \(\Pi\) take here the form
\[
\begin{align*}
J : \xi &\mapsto (X, X \circ \xi) \\
\Phi : \Xi = (\zeta_1, \zeta_2) &\mapsto U = (\zeta_1, u, v^3), \\
\Pi : U = (\zeta_1, u, v^3) &\mapsto u,
\end{align*}
\]
where \((u, v^3)\) is a solution to the system
\[
\begin{align*}
\begin{cases}
Lu^3 &= G'(u)G(u)\zeta_2 + F(u, v^3, \zeta_1) \\
u &= G(u) \prec \zeta_1 + v^3
\end{cases}
\end{align*}
\]
where \(F(u, v^3, \zeta_1)\) takes values in \(C_T^0\). The equation for \((u, v^3)\), albeit not a standard PDE, can nonetheless be solved by usual fixpoint methods, at least locally in time\(^3\).

This last example allowed us to introduce the paraprocontrolled ansatz and the use of paraproducts to describe spaces of distributions with specific behaviour. All the other examples of singular SPDEs which we mentioned in the introduction can be analysed via a change of variables involving linear combinations of paraproducts and smooth remainder terms. We illustrate the final result of the analysis instead of going step by step as we did so far. For the dynamic \(\Phi_4\) model (4) we proceed as for \(\Phi_2\) and introduce further terms by writing
\[
\varphi = X + Y + \varphi^Q, \\
\varphi^Q = \psi \prec Q + \varphi^3,
\]

\(^2\)This is not exactly true, we also need some time regularity of \(f\) but for simplicity we omit this in the discussion.

\(^3\) We pick up a quadratic estimate from the paralinearisation result (6) because it is based on a second order Taylor expansion, and therefore we cannot exclude that the solutions blows up in finite time. But given an a priori bound on the \(L^\infty\) norm of \(u\) one can show that \((u, v^3)\) stays bounded in \(C_T^1 \times C_T^1\), see [GIP15], and such an a priori bound can for certain nonlinearities \(G\) be derived from the maximum principle, see [CFG17].
where the functions \( X \in \mathcal{C}_T \mathcal{C}^{-1/2-}, Y \in \mathcal{C}_T \mathcal{C}^{1/2-}, Q \in \mathcal{C}_T \mathcal{C}^{1-}, \psi \in \mathcal{C}_T \mathcal{C}^{1/2-}, \varphi^\dagger \in \mathcal{C}_T \mathcal{C}^{3/2-} \) solve

\[
\mathcal{L} X = \zeta, \quad \mathcal{L} Y = -X^3, \quad \mathcal{L} Q = -X^2, \quad \mathcal{L} \varphi^\dagger = -3X^2 \psi - 3\psi(X^2 \circ Q) + F(\psi, \varphi^\dagger, X, Y, Q), \quad \psi = 3(Y + \varphi^Q), \tag{14}
\]

where \( F \) is a continuous function mapping into \( \mathcal{C}_T \mathcal{C}^{-1/2-} \). As before the main goal of this decomposition is to rewrite all the problematic resonant products in terms of simple expressions of the driving noise \( \xi \). The enhancement \( J \), the solution map \( \Phi \) and the projection \( \Pi \) take here the form

\[
J : \xi \mapsto (X, Y, Q, X^2 \circ Y, X^2 \circ Q) \\
\Phi : \Xi = (\zeta_1, \ldots, \zeta_5) \mapsto U = (\zeta_1, \zeta_2, \zeta_3, \psi, \varphi^\dagger) \\
\Pi : U = (\zeta_1, \zeta_2, \zeta_3, \psi, \varphi^\dagger) \mapsto \zeta_1 + \zeta_2 + \psi \leftarrow \zeta_3 + \varphi^\dagger,
\]

where the pair \((\psi, \varphi^\dagger)\) solves the equations (14) above with the driving features \((X, Y, Q, X^2 \circ Y, X^2 \circ Q)\) replaced by generic functions \(\Xi = (\zeta_1, \ldots, \zeta_5)\) with specific regularities, which in this case can be taken as

\[
\Xi \in \mathcal{C}_T \mathcal{C}^{-1/2-} \times \mathcal{C}_T \mathcal{C}^{1/2-} \times \mathcal{C}_T \mathcal{C}^{1-} \times \mathcal{C}_T \mathcal{C}^{-1/2-} \times \mathcal{C}_T \mathcal{C}^{-0-}. 
\]

The details can be found in [CC13], see also [MW16] for a proof that solutions exist for all times.

In the case of the KPZ equation (2) the decomposition is even more involved and the enhancement \( J \), the solution map \( \Phi \) and the projection \( \Pi \) take the form

\[
J : \xi \mapsto (Y, Y^\nu, Y^\varnothing, Y^\Phi, \partial_x Y \circ \partial_x JY) \\
\Phi : \Xi = (\zeta_1, \ldots, \zeta_5) \mapsto U = (\zeta_1, \zeta_2, \zeta_3, \psi, h^\dagger) \\
\Pi : U = (\zeta_1, \zeta_2, \zeta_3, \psi, h^\dagger) \mapsto h = \zeta_1 + \zeta_2 + 2\zeta_3 + \psi \leftarrow \zeta_4 + h^\dagger,
\]

where \( J(t) = \int_0^t P_{t-s} v(s) ds \) and we recall that \((P_t)_{t \geq 0}\) is the heat semigroup, and where

\[
\mathcal{L} Y = \zeta, \quad \mathcal{L} Y^\nu = (\partial_x Y)^2, \quad \mathcal{L} Y^\varnothing = \partial_x Y \partial_y Y, \quad \mathcal{L} Y^\Phi = \partial_x Y \circ \partial_x Y, \\
\mathcal{L} Y^\nu = (\partial_x Y^\nu)^2, \quad \psi = 2(\psi \leftarrow \zeta_4 + h^\dagger) + 4Y^\varnothing, \quad \mathcal{L} h^\dagger = F(\Xi, \psi, h^\dagger)
\]

for a continuous function \( F \), see [GP17]. We provide more details for the CSBE equation (3)

\[
\mathcal{L} u = \chi \partial_x u^2 + \partial_x \zeta, \tag{16}
\]

with a general constant \( \chi \) in front of the nonlinearity because this will be needed in Section 6. The change of variables reads

\[
u = X + \chi X^\nu + 2\chi^2 X^\varnothing + u^Q, \quad u' = 2u^Q + 4\chi^2 X^\varnothing, \quad u^Q = u' \leftarrow Q + u^2, \tag{17}
\]

where \( u^2 \) solves

\[
\mathcal{L} u^2 = \chi \partial_x u^2 - \chi \partial_x X^2 - 2\chi \partial_x (X^\nu X) - \mathcal{L}(u' \leftarrow Q) = \chi \partial_x (X^\nu + 2\chi^2 X^\varnothing + u^Q)^2 + 2\chi \partial_x [X(2\chi^2 X^\varnothing + u^Q)] - \mathcal{L}(u' \leftarrow Q) = \chi^3 \mathcal{L} X^\nu + 2\chi \partial_x [u^Q X - u^Q \leftarrow X] + 2\chi [\partial_x (u^Q \leftarrow X) - u^Q \leftarrow \partial_x X] + 4\chi^3 [\partial_x (X^\varnothing \leftarrow X) + 4\chi^3 [\partial_x (X^\varnothing \leftarrow X) - X^\varnothing \leftarrow \partial_x X] + \chi \partial_x [2\chi X^\nu (2\chi^2 X^\varnothing + u^Q) + (2\chi^2 X^\varnothing + u^Q)^2] - [\mathcal{L}(u' \leftarrow Q) - u' \leftarrow (\mathcal{L} Q)],
\]

11
and the enhanced noise is defined by
\[ \Xi = J(\xi) = (X, X, X, X, X, X, Q, Q \circ X), \]
where \( L_X = \partial_x \xi, \quad L_Q = \partial_x X, \)
and
\[ L_{X^2} = \partial_x (X^2), \quad L_{XX} = \partial_x (XX^\circ X), \quad L_{X^\circ} = \partial_x (X^\circ)^2. \]

4 Ambiguities and renormalisation

The previous analysis reduces the study of a singular SPDE to that of the enhancement \( J \) and of the enhanced solution map \( \Phi \). The enhanced version of the singular equation is a standard PDE for a new unknown together with some paradifferential relations. This factorisation “distills” in the definition of the enhancement \( \Xi = J(\xi) \) all the problematic products (or resonant products). One cannot expect to be able to analyse in full generality the map \( J \) without using specific properties of the driving function \( \xi \). Two basic difficulties are:

1. the lack of continuity of the resonant products implies that these products are essentially undefined and the enhancement map \( J \) can be extended to irregular inputs \( \xi \) only within a specific approximation procedure (or not at all);
2. for most of the above examples, even within the more restricted context where we only try to extend \( J \) to a given stochastic process \( \xi \) through a specific approximation procedure, the enhancement map \( J \) fails to extend due to divergences: the only natural limit of the resonant products is infinite.

The first difficulty means that any resonant product which is formed without sufficient regularity should be considered inherently ambiguous, that is non-robust to different approximation procedures. A satisfying analysis of such ambiguities is still lacking in the paracontrolled setting and much more developed in the setting of regularity structures [BHZ16] where deep connection with algebra and renormalization procedures in Quantum Field Theory have been pointed out. For some details we suggest the reader to have a look at L. Zambotti’s contribution in this volume. As an example we take here the case of the solution theory for the gPAM (13). In order to describe the effects of the ambiguity on the equation we construct an extended solution map \( \Psi_{\text{ext}} \) for smooth inputs \( \eta \) by translations \( T_C J(\eta) = (\eta, (J(\eta)) \circ \eta + C) \) of the enhanced gPAM noise \( J(\eta) \) (here we denote as usual with \( J(\eta) \) a suitable solution to the parabolic problem \( L J(\eta) = \eta \)). Setting
\[ \hat{u} = \Psi_{\text{ext}}(\xi, C) := \Pi \circ \Phi(T_C J(\eta)), \]
the reader can easily check that the function \( \hat{u} \) satisfies a modified PDE which reads
\[ L \hat{u} = G(\hat{u}) \eta + G'(\hat{u}) G(\hat{u}) C. \quad (18) \]
But given \( \eta \in C^\infty(\mathbb{T}^2) \) and \( C > 0 \) one can find a sequence of smooth functions \( (\eta_n) \subset C^\infty(\mathbb{T}^2) \) such that \( \eta_n \) converges to \( \eta \) in \( C^{-1-} \) but \( J(\eta_n) \) converges to \( T_C J(\eta) \). So the analysis of the gPAM model (5) together with the requirement of stability under perturbations which are only small in very weak topologies generates quite naturally a relaxed equation of the form (18). Similar considerations are applicable, at least in principle, to all the other singular SPDEs.

The problem of renormalisation is related to this ambiguity and to the robustness of the form of the equations under irregular perturbations. Here the problem is however that certain products are intrinsically impossible to define due to the presence of infinities and that some subtraction is required for them to have a finite limit. One of the simplest
situations is still that of the 2d gPAM driven by space white noise. In this case the product
\(X \circ \xi\), understood as limit of smooth convolutional approximations is almost surely infinite.

So if want the equations with mollified noise to have a well defined limit, then we need
to start with a suitably modified equation of the form given by eq. (18) and conceived in
such a way that the additional term provides the necessary cancellations to remove the
divergence in the resonant product. Denoting by \(\xi\) the convolutional regularisation of
the white noise \(\xi\) and by \(c_\varepsilon\) a family of constants we see that if \(u_\varepsilon\) is the solution to
\[
\begin{align*}
L u_\varepsilon &= G(u_\varepsilon) \xi - G'(u_\varepsilon) G(u_\varepsilon) \xi, \\
\text{then } \hat{u}_\varepsilon &= \Psi_{\text{ext}}(\xi_\varepsilon, -c_\varepsilon) = \Phi(T_{-c_\varepsilon} J(\eta)) = \Phi((\xi_\varepsilon, J_{\varepsilon} \xi_\varepsilon - c_\varepsilon)).
\end{align*}
\]
In other words \(\hat{u}_\varepsilon\) is a continuous function of the quantities \(\xi_\varepsilon, J_{\varepsilon} \xi_\varepsilon - c_\varepsilon\). A similar result is also
true for the right hand side of the equation, namely
\[
G(u_\varepsilon) \circ \xi_\varepsilon := G(\hat{u}_\varepsilon) \xi - G'(\hat{u}_\varepsilon) G(\hat{u}_\varepsilon) c_\varepsilon.
\]

By a probabilistic analysis it can be checked that a (non-unique) choice of \((c_\varepsilon)\) exists for
which \((\Xi_\varepsilon)\) converges (in probability and in \(L^p\) with respect to the randomness) in
the appropriate topology. Denoting by \(\Xi\) the limit we have that also the solution converges
\(\hat{u}_\varepsilon \to \hat{u}\) and satisfies the renormalised singular SPDE
\[
L \hat{u} = G(\hat{u}) \circ \Xi,
\]
where the right hand side now is a certain non-linear function of \(\hat{u}\) and \(\Xi\) which can be
identified with the limit \(G(\hat{u}) \circ \Xi = \lim_{\varepsilon \to 0} G(\hat{u}_\varepsilon) \circ \Xi_\varepsilon\). In this particular case the limit is
controlled via the parabolised Ansatz for \(u_\varepsilon\) and via the continuity results which follow
from it and from the convergence of the enhanced noise \(\Xi_\varepsilon\).

More complex renormalisations are necessary in other equations. For example, in the
case of the \(\Phi^4_3\) model the ill-defined products are
\[
X^2, X^3, X^2 \circ J(X^3), X^2 \circ J(X^2),
\]
see eq. (15), where \(X = J(\xi)\). Stochastic analysis shows that there exists a choice of constants
\((c_1, c_2) > 0\) such that \(c_1^2, c_2^2 \to \infty\) as \(\varepsilon \to 0\) for which the convergence in probability
\[
\begin{align*}
X_\varepsilon^{\circ 3} &:= X_\varepsilon^3 - 3c_1^2 \varepsilon X_\varepsilon \to X^{\circ 3}, \\
X_\varepsilon^{\circ 2} &:= X_\varepsilon^2 - c_2^2 X_\varepsilon \to X^{\circ 2}, \\
X_\varepsilon^{\circ 2} \circ J(X_\varepsilon^{\circ 3}) &:= X_\varepsilon^{\circ 2} \circ J((X_\varepsilon^{\circ 3}) - 3c_1^2 X_\varepsilon \to X^{\circ 2} \circ J(X^{\circ 3}), \\
X_\varepsilon^{\circ 2} \circ J(X_\varepsilon^{\circ 2}) &:= X_\varepsilon^{\circ 2} \circ J(X_\varepsilon^{\circ 2}) - c_2^2 \varepsilon \to X^{\circ 2} \circ J(X^{\circ 2})
\end{align*}
\]
holds in the appropriate spaces, where \(X_\varepsilon = \rho_\varepsilon * X\) is a convolutional regularisation of \(X\) with a smooth kernel. As a consequence of these convergence results and of the structure
of the solution theory described by eq. (15) the function \(\varphi_\varepsilon = \Pi \circ \Phi(\Xi_\varepsilon)\), where
\[
\Xi_\varepsilon = (X_\varepsilon, J(X_\varepsilon^{\circ 3}), J(X_\varepsilon^{\circ 2}), X_\varepsilon^{\circ 2} \circ J(X_\varepsilon^{\circ 3}), X_\varepsilon^{\circ 2} \circ J(X_\varepsilon^{\circ 2})),
\]
solves the renormalised \(\Phi^4_3\) equation
\[
\begin{align*}
L \varphi_\varepsilon &= -\varphi_\varepsilon^3 + 3(c_1^2 + c_2^2) \varphi_\varepsilon + \xi_\varepsilon
\end{align*}
\]
and converges to \(\varphi = \Pi \circ \Phi(\Xi)\) where \(\Xi := \lim_\varepsilon \Xi_\varepsilon\) which again can be described in terms
of a limiting renormalised singular SPDE. In this review we will not discuss details of
the convergence results \(\Xi_\varepsilon \to \Xi\). These are mostly handled with standard probabilistic
techniques. A very comprehensive convergence theory for the stochastic terms exists in
terms of regularity structures [BHZ16, CH16] but most of the ideas can be employed also
within the paracontrolled approach, see also [MWX16].
5 Higher order expansions

The theory of paracontrolled distributions that we discussed so far is essentially a first order calculus. For example, in the parabolic Anderson model (gPAM with \( G(u) = u \)) we expand the solution in terms of a paraproduct \( u \ll X \) plus a smoother remainder \( u^\dagger \), where \( \mathcal{L}X = \xi \) and \( X \in C_TC^\alpha \), \( u^\dagger \in C_TC^\alpha \) for \( \alpha = 1 \). Then terms like the resonant product \( u^\dagger \circ \xi \) pose no problem because \( \xi \in C^{\alpha - 2} \) and \( u^\dagger \in C_TC^{2\alpha} \) and the sum of the regularities is \( 3\alpha - 2 > 0 \). But what if \( \xi \) has worse regularity, so if \( \alpha \leq 2/3 \) and therefore \( 3\alpha - 2 \leq 0 \)?

One relevant example is the parabolic Anderson model in \( d = 3 \), where the space white noise has regularity strictly less than \(-3/2\) and therefore \( \alpha < 1/2 \). The idea, inspired by (controlled) rough paths, is then to find a higher order expansion of \( u \) which should be of the type

\[
u = \sum_{k=1}^{n-1} \sum_{\tau \in \mathcal{I}_k} u^\tau \ll X^\tau + u^\dagger \]

for some \( u^\dagger \in C_TC^{\alpha_0} \) and suitable \( (X^\tau)_{\tau \in \mathcal{I}_k} \), with \( X^\tau \in C_TC^{\alpha_0} \) for all \( \tau \in \mathcal{I}_k \). If \( (n+1)\alpha - 2 > 0 \), then at least the product \( u^\dagger \circ \xi \) is well defined and this gives us some hope to construct a continuous map \( u, (u^\tau, \tau, u^\dagger, \xi, (X^\tau)_\tau, \ldots) \mapsto u\xi \). But what should the \( (X^\tau) \) be and how do we define the map? This is not very obvious and for quite some time the extension of paracontrolled distributions to more irregular driving noises remained open. But recently Bailleul and Bernicot [BB16b] have made significant progress in that direction.

To understand their results let us have a look at the parabolic Anderson model

\[ \mathcal{L}u = u\xi \]

for \( \xi \in C^{\alpha - 2} \) with \( \alpha \in (1/2, 2/3) \), so slightly better than the white noise in \( d = 3 \) but worse than the white noise in \( d = 2 \). We guess the expansion \( u = \sum_{k=1}^{n-1} \sum_{\tau \in \mathcal{I}_k} u^\tau \ll X^\tau + u^\dagger \) with \( u^\dagger \in C_TC^{3\alpha} \) and \( (u^\tau) \) and \( (X^\tau) \) to be determined. Then

\[
u\xi = u\ll \xi + u\ll \xi + \left( \sum_{k=1}^{n-1} \sum_{\tau \in \mathcal{I}_k} u^\tau \ll X^\tau \right) \circ \xi + u^\dagger \circ \xi,
\]

where as before we single out the problematic resonant product with the underbrace “!!”: If we assume that \( u^\tau \in C_TC^{\alpha} \) for all \( \tau \in \mathcal{I}_1 \cup \mathcal{I}_2 \) and \( X^\tau \in C_TC^{\alpha} \) for all \( \tau \in \mathcal{I}_k \), then for \( \tau \in \mathcal{I}_2 \) the resonant product \( (u^\tau \ll X^\tau) \circ \xi \) can be controlled with the commutator estimate:

\[
(u^\tau \ll X^\tau) \circ \xi = (C(u^\tau, X^\tau, \xi) + u^\tau (X^\tau \circ \xi)),
\]

where we used that \( \alpha < 1 \) and \( 4\alpha - 2 > 0 \) to see that the commutator is bounded, and where we need to assume that \( X^\tau \circ \xi \in C_TC^{3\alpha - 2} \) is extrinsically given. However, for \( \tau \in \mathcal{I}_1 \) the term \( C(u^\tau, X^\tau, \xi) \) is still not well defined because the sum of the regularities is \( 3\alpha - 2 < 0 \). To deal with the resonant product \( (u^\tau \ll X^\tau) \circ \xi \) we therefore assume that \( u^\tau \) is itself paracontrolled of order \( 2\alpha \) for all \( \tau \in \mathcal{I}_1 \):

\[
u^\tau = \sum_{\tau' \in \mathcal{I}_1} u^{\tau, \tau'} \ll X^{\tau'} + u^{\tau, \dagger}
\]

with \( u^{\tau, \tau'} \in C_TC^\alpha \) and \( u^{\tau, \dagger} \in C_TC^{2\alpha} \). Then

\[
(u^{\tau, \dagger} \ll X^\tau) \circ \xi = C(u^{\tau, \dagger}, X^\tau, \xi) + u^{\tau, \dagger}(X^\tau \circ \xi).
\]
At this point we would like to gain 2α degrees of regularity from \( u^{r,d} \) in the commutator to see that it is in \( C_T C^{2\alpha-2} \), but this is not possible because 2α > 1 and the commutator estimate Lemma 2.1 allows us only to gain less than one derivative. However, the sum of the regularities of \( X^\tau \) and \( \xi \) is 2α − 2 > −1, and therefore we can use that \( u^{r,d} \in C_T C^{1-} \) to obtain

\[
\|C(u^{r,d}, X^\tau, \xi)\|_{C_T C^{1(--) + 2\alpha - 2}} \lesssim \|u^{r,d}\|_{C_T C^{1-}} \|X^\tau\|_{C_T C^{0}} \|\xi\|_{C_T C^{0-2}} \\
\lesssim \|u^{r,d}\|_{C_T C^{2\alpha}} \|X^\tau\|_{C_T C^{\alpha}} \|\xi\|_{C_T C^{\alpha-2}}.
\]

Since we estimate the commutator in a space of positive regularity and 3α − 2 < 0, we still get \( C(u^{r,d}, X^\tau, \xi) \in C_T C^{3\alpha-2} \).

**Remark 5.1** This argument is particular to the not-so-singular problem studied here and breaks down if \( \alpha < 1/2 \). In that case it may be necessary to develop a version of the commutator estimate which allows to gain more than one derivative from \( u^{r,d} \). This can be achieved by subtracting not only \( u^{r,d} (X^\tau \circ \xi) \) from \( (u^{r,d} \prec X^\tau) \circ \xi \) but also further correction terms that involve modified Littlewood-Paley blocks and roughly speaking correspond to polynomial terms in regularity structures. Currently there is no reference where this is worked out.

Next, we note that the product \( u^{r,d} (X^\tau \circ \xi) \) is well defined if \( X^\tau \circ \xi \in C_T C^{2\alpha-2} \) is given because then the sum of the regularities of its factors is 4α − 2 > 0. It remains to understand the resonant product

\[
((u^{r,r'} \prec X^\tau') \prec X^\tau) \circ \xi = C(u^{r,r'} \prec X^\tau', X^\tau, \xi) + (u^{r,r'} \prec X^\tau') (X^\tau \circ \xi)
\]

where our commutator estimate really fails: After all \( u^{r,r'} \prec X^\tau' \in C_T C^\alpha, X^\tau \in C_T C^\alpha, \) and \( \xi \in C_T C^{\alpha-2} \) so the sum of the regularities is 3α − 2 < 0. But Bailleul and Bernicot realised that one can iterate the commutator estimate in the following way:

**Lemma 5.2 ([BB16b], formula(3.8))** Let \( \alpha, \beta, \gamma, \delta \in \mathbb{R} \) be exponents such that \( \alpha + \beta + \gamma + \delta > 0 \), then there exists a four-linear map \( C^{(2)} : C^\alpha \times C^\beta \times C^\gamma \times C^\delta \to C^{\alpha + \beta + \gamma + \delta} \) such that if \( f, g, h, \xi \) are smooth functions we have

\[
C^{(2)}(f, g, h, \xi) = C(f \prec g, h, \xi) - f C(g, h, \xi).
\]

Therefore, we can set

\[
C(u^{r,r'} \prec X^\tau', X^\tau, \xi) = C^{(2)}(u^{r,r'}, X^\tau', X^\tau, \xi) + u^{r,r'} C(X^\tau', X^\tau, \xi)
\]

which is well defined provided that \( C(X^\tau', X^\tau, \xi) \) is given and has its natural regularity \( C_T C^{3\alpha-2} \). The only remaining product that we still need to control is then

\[
(u^{r,r'} \prec X^\tau')(X^\tau \circ \xi) = (u^{r,r'} \prec X^\tau') (X^\tau \circ \xi) + (u^{r,r'} \prec X^\tau') (X^\tau \circ \xi)
\]

\[
+ C(u^{r,r'}, X^\tau', X^\tau \circ \xi) + u^{r,r'} (X^\tau \circ (X^\tau \circ \xi)),
\]

which is under control as long as \( X^\tau \circ (X^\tau \circ \xi) \in C_T C^{3\alpha-2} \) is given. So let us put everything together:
\[
\begin{align*}
\xi u = \xi \overline{u} + \xi \times \alpha \dot{u} + \xi \overline{\alpha} \dot{u} + \sum_{\tau \in \mathcal{I}_2} (C(u^\tau, X^\tau, \xi) + u^\tau(X^\tau \circ \xi)) \\
+ \sum_{\tau \in \mathcal{I}_1} (C(u^\tau \times \tau, X^\tau, \xi) + u^\tau(X^\tau \circ \xi)) \\
+ \sum_{\tau, \tau' \in \mathcal{I}_1} (C(\tau, \tau', X^\tau, X^\tau, \xi)) u^\tau \tau' (X^\tau \circ \tau) \\
+ \sum_{\tau, \tau' \in \mathcal{I}_1} ((u^\tau \tau', \dot{X}^\tau) \circ (X^\tau \circ \tau) + (u^{\tau'} \tau', \dot{X}^{\tau'}) \circ (X^\tau \circ \tau)),
\end{align*}
\]

where the underbrace indicates the regularity of each term. Therefore, the product \(u \xi\) is under control if all of the following terms are extrinsically given

\[
\{X^\sigma \circ \xi, X^\sigma \circ \xi, C(X^\tau, X^\tau, \xi), X^\tau \circ \tau(\tau \circ \sigma) : \tau, \tau' \in \mathcal{I}_1, \sigma \in \mathcal{I}_2\}
\]

and have their natural regularity.

But making sense of the product \(u \xi\) is only the first step in the analysis of the equation. Next, we should check that for a paracontrolled \(u\) also the solution \(v\) to \(Lv = u \xi\) is paracontrolled. Here a problem occurs: above we discussed that if we make the paracontrolled Ansatz

\[
v = \sum_{k=1}^{2} \sum_{\tau \in \mathcal{I}_k} v^\tau \otimes X^\tau + u^\tau
\]

(now with the modified paraproduct \(\otimes\) rather than \(\triangleleft\)), then we can commute the operator \(\mathcal{L}\) with \(\otimes\) in the sense that \(\mathcal{L}(v^\tau \otimes X^\tau) - v^\tau \otimes \mathcal{L}X^\tau\) has higher regularity. However, the commutation can gain at most one degree of regularity from \(v^\tau\), and as we have just seen in Remark 5.1 this may not always be sufficient. So Bailleul and Bernicot introduce an “intertwined” paraproduct defined as

\[
(f \otimes g)(t) = \int_0^t P_t \triangleleft (f \triangleleft Lg)(s) ds,
\]

where \((P_t)_{t \geq 0}\) is the heat semigroup generated by \(\Delta\). Then by definition the exact relation \(\mathcal{L}(f \otimes g) = f \triangleleft Lg\) holds, without error term. So we make the modified paracontrolled Ansatz

\[
v = \sum_{k=1}^{2} \sum_{\tau \in \mathcal{I}_k} v^\tau \otimes X^\tau + u^\tau
\]

with the same regularities as above. Then

\[
\mathcal{L}v = \sum_{k=1}^{2} \sum_{\tau \in \mathcal{I}_k} v^\tau \otimes \mathcal{L}X^\tau + \mathcal{L}u^\tau,
\]

and if we take \(\tau_1 \in \mathcal{I}_1\) with \(\mathcal{L}X_\tau = \xi\) and \(v_\tau = \xi\), then \(v_\tau \otimes \mathcal{L}X_\tau\) cancels the worst regularity contribution \(u \otimes \xi \in C_\tau \mathcal{C}^\alpha\) to \(u \xi\). The remaining \(X^\tau\) and \(v^\tau\) have to be chosen such that all contributions of regularity \(2\alpha - 2\) cancel. The most tricky term to deal with is \(u \otimes \xi = \xi \otimes u\) which we decompose as

\[
\xi \otimes u = \xi \otimes \left(\sum_{k=1}^{2} \sum_{\tau \in \mathcal{I}_k} u^\tau \otimes X^\tau + u^\tau\right) = \sum_{\tau \in \mathcal{I}_1} \xi \otimes (u^\tau \otimes X^\tau) + \sum_{\tau \in \mathcal{I}_2} \xi \otimes (u^\tau \otimes X^\tau) + \xi \otimes u^\tau.
\]

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The first sum on the right hand side still has regularity $2\alpha - 2$, but Theorem 8 of [BB16b] gives

$$ T(\xi, u^\tau, X^\tau) = \xi \langle u^\tau \lhd X^\tau \rangle - u^\tau \langle \xi \lhd X^\tau \rangle \in C_T C^{3\alpha - 2}. $$

Thus, we have

$$ \xi \langle u = \sum_{\tau \in \mathcal{I}_1} T(\xi, u^\tau, X^\tau) + \sum_{\tau \in \mathcal{I}_2} \xi \langle u^\tau \lhd X^\tau \rangle + \xi \langle u^\tau \rangle + \sum_{\tau \in \mathcal{I}_1} u^\tau \langle \xi \lhd X^\tau \rangle, $$

and we need a contribution from $\mathcal{L}v$ to cancel the last term on the right hand side. The other terms in the product $u\xi$ which have regularity worse than $3\alpha - 2$ are $\sum_{\tau \in \mathcal{I}_1} u^\tau \lhd (X^\tau \circ \xi)$ and $\sum_{\tau, \tau' \in \mathcal{I}_1} (u^\tau, u^{\tau'}) \langle (X^\tau \circ \xi) \rangle$, and for fixed $\tau \in \mathcal{I}_1$ we have

$$ u^\tau \lhd (X^\tau \circ \xi) - u^\tau \lhd (X^\tau \circ \xi) \in C_T C^{3\alpha - 2}, $$

so overall the contribution of regularity $2\alpha - 2$ is

$$ \sum_{\tau \in \mathcal{I}_1} \left(u^\tau \lhd (X^\tau \circ \xi) + \sum_{\tau' \in \mathcal{I}_1} (u^\tau, u^{\tau'}) \langle (X^\tau \circ \xi) \rangle \right) = \sum_{\tau \in \mathcal{I}_1} u^\tau \langle (X^\tau \circ \xi). $$

In conclusion, we obtain the decomposition

$$ u\xi - u \langle \xi - \sum_{\tau \in \mathcal{I}_1} u^\tau \langle \{\xi \lhd X^\tau \} + (X^\tau \circ \xi) \rangle \in C_T C^{3\alpha - 2}. $$

We can rewrite $X^\tau \circ \xi + \xi \lhd X^\tau = X^\tau \xi - X^\tau \lhd \xi$, and therefore we should set $\mathcal{I}_1 = \{1\}$ with $LX^1 = \xi$ and $\mathcal{I}_2 = \{2\}$ with $LX^2 = X^1 \xi - X^1 \lhd \xi$. Then we get with the Ansatz $v^1 = u$ and $v^2 = u^1$ that

$$ \mathcal{L}v = u \langle \xi + u^1 \lhd (X^1 \xi - X^1 \lhd \xi) \rangle + \mathcal{L}v^2, $$

so if we set $\mathcal{L}v = u\xi$, then

$$ \mathcal{L}v^2 = u\xi - u \langle \xi - u^1 \lhd (X^1 \xi - X^1 \lhd \xi) \rangle \in C_T C^{3\alpha - 2}, $$

which proves that $v^2 \in C_T C^{3\alpha}$ and therefore the paracontrolled Ansatz was justified. Moreover, since $v^1 = u$ we have with $v^{1,1} = u^1$ that $v^1 - v^{1,1} \lhd X^1 = u - u^1 \lhd X^1 \in C_T C^{2\alpha}$ and therefore also $v^1$ is paracontrolled. The terms in (19) that we need to construct in order to make sense of all the products are

$$ X^1 \circ \xi, X^2 \circ \xi, C(X^1, X^1, \xi), X^1 \circ (X^1 \circ \xi). $$

But of course now we were inconsistent: We started with $u = u^1 \lhd X^1 + u^2 \lhd X^2 + u^5$ and ended up with a $v$ that on the first level is paracontrolled in terms of the new intertwined paraproduct,

$$ v - v^1 \lhd X^1 - v^2 \lhd X^2 \in C_T C^{3\alpha}, $$

but on the second level is paracontrolled in terms of the usual paraproduct, $v^1 - v^{1,1} \lhd X^1 \in C_T C^{2\alpha}$. To set up a Picard iteration the map that sends $u$ to the solution $v$ of $\mathcal{L}v = u\xi$ should map the space of paracontrolled distributions into itself, so we should assume that also $u$ was paracontrolled in terms of $\ast \lhd \ast$ and also $u^1$ is paracontrolled in terms of $\ast \lhd \ast$. For that we need to understand the relation between the two paraproducts $\ast \lhd \ast$ and $\ast \lhd \ast$ and also some commutator estimates involving $\ast \lhd \ast$. All this is worked out in [BB16b], where also the nonlinear case of gPAM with $\xi$ of regularity $\xi \in C^{\alpha - 2}$ with $\alpha < 2/3$ is treated. In that case we also need a higher order version of the paralinearisation result (6), but this is relatively easy to derive.
6 Weak universality

One prominent application of the theory of singular SPDEs is the derivation of scaling limits of random fields described by local non-linear stochastic dynamics. As an example we will sketch the case of the CSBE equation which has been first analysed via regularity structures by Hairer and Quastel [HQ15]. As a mesoscopic model of a weakly-asymmetric diffusion we will consider the solution of the following SPDE. Take a small \( \varepsilon > 0 \) and let \( \mathbb{T}_\varepsilon = \mathbb{T}/\varepsilon \) and \( \nu : \mathbb{R}_+ \times \mathbb{T}_\varepsilon \to \mathbb{R} \) be the solution to

\[
\mathcal{L} \nu = \varepsilon^{1/2} \partial_x P(\nu) + \partial_x \eta
\]

where \( \eta \) is a Gaussian noise on \( \mathbb{R}_+ \times \mathbb{T}_\varepsilon \) with finite-range space-time correlations and \( P : \mathbb{R} \to \mathbb{R} \) is a given smooth function. We assume that \( \eta \) is centred with covariance

\[
\mathbb{E}[\eta(t,x)\eta(s,y)] = C_\varepsilon(t-s,x-y), \quad t, s \in \mathbb{R}, x, y \in \mathbb{T}_\varepsilon,
\]

where \( C_\varepsilon \) is (in the second variable) the \( \mathbb{T}_\varepsilon \)-periodised version of a function \( C : \mathbb{R} \times \mathbb{R} \to \mathbb{R} \) which has with sufficient polynomial decay in both variables. The parabolic change of variables \( v_\varepsilon(t,x) = \varepsilon^{-1/2} \nu(\varepsilon x^2, x/\varepsilon) \) gives the equation

\[
\mathcal{L} v_\varepsilon = \varepsilon^{-1} \partial_x P(\varepsilon^{1/2} v_\varepsilon) + \partial_x \xi_\varepsilon
\]

where \( \xi_\varepsilon = \varepsilon^{-3/2} \eta(\cdot/\varepsilon^2, \cdot/\varepsilon) \) is a noise which converges to a space-time white noise \( \xi \). Indeed the rescaled fields \( v_\varepsilon, \xi_\varepsilon \) live on the standard torus \( \mathbb{T} \) and

\[
\mathbb{E}[\xi_\varepsilon(t,x)\xi_\varepsilon(s,y)] = \varepsilon^{-3} C_\varepsilon((t-s)/\varepsilon^2, (x-y)/\varepsilon), \quad t, s \in \mathbb{R}, x, y \in \mathbb{T},
\]

where, as \( \varepsilon \to 0 \) we have \( \varepsilon^{-3} C_\varepsilon((t-s)/\varepsilon^2, (x-y)/\varepsilon) \to \delta(t-s)\delta(x-y) \) weakly as a space-time distribution. The goal of the analysis is to show is that as \( \varepsilon \to 0 \) the function \( v_\varepsilon \) converges to the solution of the CSBE equation (3) with a specific constant \( \chi \) in front of the non-linearity. The constant will depend only on the shape of the function \( P \). By going to a reference frame via a constant velocity \( a_\varepsilon \) change of variables we have that \( v_\varepsilon = v_\varepsilon(t,x) = \varepsilon^{-1/2} \nu(t/\varepsilon^2, (x+a_\varepsilon t)/\varepsilon) \) is the solution to

\[
\mathcal{L} v_\varepsilon = a_\varepsilon \partial_x v_\varepsilon + \varepsilon^{-1} \partial_x P(\varepsilon^{1/2} v_\varepsilon) + \partial_x \xi_\varepsilon = \varepsilon^{-1} \partial_x \tilde{P}(\varepsilon^{1/2} v_\varepsilon) + \partial_x \xi_\varepsilon,
\]

where \( \tilde{P}(x) = P(x) + a_\varepsilon \varepsilon^{1/2} x \). We make an Ansatz of the form

\[
v_\varepsilon = X\varepsilon + \chi \tilde{X}_\varepsilon + 2\chi^2 \tilde{X}_\varepsilon^2 + \nu_\varepsilon^Q,
\]

where \( \chi \) is a real number different from 0 and \( \tilde{X}_\varepsilon \) and \( \tilde{X}_\varepsilon^2 \) are functions, all to be determined later. We then use a Taylor expansion to get

\[
\frac{1}{\varepsilon} \tilde{P}(\varepsilon^{1/2} v_\varepsilon) = \frac{1}{\varepsilon} \tilde{P}(\varepsilon^{1/2} X\varepsilon) + \frac{1}{\varepsilon^{1/2}} \tilde{P}'(\varepsilon^{1/2} X\varepsilon)(\chi \tilde{X}_\varepsilon + 2\chi^2 \tilde{X}_\varepsilon^2 + \nu_\varepsilon^Q) + \frac{1}{2} \tilde{P}''(\varepsilon^{1/2} X\varepsilon)(\chi \tilde{X}_\varepsilon + 2\chi^2 \tilde{X}_\varepsilon^2 + \nu_\varepsilon^Q)^2 + R_\varepsilon,
\]

where

\[
R_\varepsilon = \frac{\varepsilon^{1/2}}{2} \int_0^1 d\tau (1-\tau)^2 \tilde{P}^{(3)}(\varepsilon^{1/2} X\varepsilon + \tau \varepsilon^{1/2}(\chi \tilde{X}_\varepsilon + 2\chi^2 \tilde{X}_\varepsilon^2 + \nu_\varepsilon^Q))(\chi \tilde{X}_\varepsilon + 2\chi^2 \tilde{X}_\varepsilon^2 + \nu_\varepsilon^Q)^3.
\]

Let assume that \( \tilde{P}^{(3)}(x) = P^{(3)}(x) \) has polynomial growth of order \( M \) and that we have the following bounds

\[
\|X\varepsilon\|_{L^\infty} \lesssim \varepsilon^{-1/2-\kappa}, \quad \|\chi \tilde{X}_\varepsilon^2 + 2\chi^2 \tilde{X}_\varepsilon^2 + \nu_\varepsilon^Q\|_{L^\infty} \lesssim \varepsilon^{-\kappa}
\]

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for some small $\kappa > 0$. Then

$$\|R_\varepsilon\|_{L^\infty} \lesssim \varepsilon^{1/2} \varepsilon^{-\kappa M} \varepsilon^{-3\kappa} \lesssim \varepsilon^{1/2-\kappa(M+3)}$$

so if $\kappa$ is small enough this remainder goes to zero in $L^\infty$. This justifies the Taylor expansion at least under the assumptions we made.

Now we set

$$\chi_\varepsilon = \tilde{P}'(\varepsilon^{1/2}X_\varepsilon), \quad \mathcal{L}X_\varepsilon = \partial_x \xi_\varepsilon, \quad 2\chi \tilde{X}_\varepsilon = \frac{1}{\varepsilon^{1/2}} \tilde{P}'(\varepsilon^{1/2}X_\varepsilon), \quad \mathcal{L}\tilde{Q}_\varepsilon = \partial_x \tilde{X}_\varepsilon, \quad \chi\mathcal{L}\tilde{X}_\varepsilon = \frac{1}{\varepsilon} \mathcal{L} \tilde{P}(\varepsilon^{1/2}X_\varepsilon),$$

and

$$\mathcal{L}\tilde{X}_\varepsilon = \partial_x (\tilde{X}_\varepsilon \circ \tilde{X}_\varepsilon), \quad \mathcal{L}\tilde{X}_\varepsilon^\# = \partial_x (\tilde{X}_\varepsilon \circ \tilde{X}_\varepsilon), \quad \chi\mathcal{L}\tilde{X}_\varepsilon^\# = \frac{1}{2} \partial_x (\chi_\varepsilon (\tilde{X}_\varepsilon\varepsilon)^2).$$

Note that $\chi$ is a real number and $\chi_\varepsilon$ is a function and the two do not agree. The reason for the notation is that in the end $\chi_\varepsilon$ will converge to $\chi$. With these definitions at hand we see that setting

$$v_\varepsilon = X_\varepsilon + \chi\tilde{X}_\varepsilon^\# + 2\chi^2 \tilde{X}_\varepsilon + v_\varepsilon^Q, \quad v'_\varepsilon = 2\chi v_\varepsilon^Q + 4\chi^3 \tilde{X}_\varepsilon \tilde{X}_\varepsilon^\#, \quad v_\varepsilon^Q = v'_\varepsilon \ll \tilde{Q}_\varepsilon + v_\varepsilon^\#$$

we have

$$\mathcal{L} v_\varepsilon^2 = \frac{1}{\varepsilon} \partial_x [\tilde{P}(\varepsilon^{1/2}v_\varepsilon)] - \frac{1}{\varepsilon} \partial_x [\tilde{P}(\varepsilon^{1/2}X_\varepsilon)] - 2\chi^2 \partial_x [(\chi\tilde{X}_\varepsilon \tilde{X}_\varepsilon^\#)] - \mathcal{L} (v'_\varepsilon \ll \tilde{Q}_\varepsilon)$$

$$= \frac{1}{\varepsilon^{1/2}} \partial_x [\tilde{P}'(\varepsilon^{1/2}X_\varepsilon)(2\chi^2 \tilde{X}_\varepsilon + v_\varepsilon^Q)] + \frac{1}{2} \partial_x [\tilde{P}''(\varepsilon^{1/2}X_\varepsilon)(\chi\tilde{X}_\varepsilon \tilde{X}_\varepsilon^\# + 2\chi^2 \tilde{X}_\varepsilon^\# + v_\varepsilon^Q)^2]$$

$$+ \partial_x R_\varepsilon - \mathcal{L} (v'_\varepsilon \ll \tilde{Q}_\varepsilon)$$

$$= \chi 2\partial_x [\tilde{X}_\varepsilon (2\chi^2 \tilde{X}_\varepsilon + v_\varepsilon^Q)] + \frac{1}{2} \partial_x [\chi_\varepsilon(\chi\tilde{X}_\varepsilon \tilde{X}_\varepsilon^\# + 2\chi^2 \tilde{X}_\varepsilon^\# + v_\varepsilon^Q)^2] + \partial_x R_\varepsilon - \mathcal{L} (v'_\varepsilon \ll \tilde{Q}_\varepsilon)$$

$$= \chi^3 \mathcal{L}\tilde{X}_\varepsilon \tilde{X}_\varepsilon^\# + 2\chi^3 \partial_x [\tilde{X}_\varepsilon \tilde{X}_\varepsilon^\# - v_\varepsilon^Q \ll \tilde{X}_\varepsilon] + 2\chi^3 [\partial_x (v_\varepsilon^Q \ll \tilde{X}_\varepsilon) - v_\varepsilon^Q \ll \partial_x \tilde{X}_\varepsilon]$$

$$+ 4\chi^3 \mathcal{L}\tilde{X}_\varepsilon \tilde{X}_\varepsilon^\# + 4\chi^3 \partial_x [\tilde{X}_\varepsilon \tilde{X}_\varepsilon^\# - \tilde{X}_\varepsilon \ll \partial_x \tilde{X}_\varepsilon]$$

$$+ \partial_x \chi_\varepsilon(2\chi\tilde{X}_\varepsilon^\# (2\chi^2 \tilde{X}_\varepsilon + v_\varepsilon^Q) + (2\chi^2 \tilde{X}_\varepsilon^\# + v_\varepsilon^Q)^2] - \mathcal{L} (v'_\varepsilon \ll \tilde{Q} - v'_\varepsilon \ll (\mathcal{L}\tilde{Q}))$$

$$+ \partial_x R_\varepsilon,$$

which has to be compared with the expansion that we used for the solution to the CSBE (17). The only structural difference is the presence of random fields $\chi_\varepsilon$ in the place of some of the $\chi$ and the additional source term $\partial_x R_\varepsilon$ which however goes to zero in a topology that is compatible with the required regularity of $v^Q_\varepsilon$. The role of the enhancement $\Xi$ is taken by the (slightly modified) family of random fields

$$\tilde{\Xi}_\varepsilon = (\chi_\varepsilon, X_\varepsilon, \tilde{X}_\varepsilon, \tilde{X}_\varepsilon^\#, \tilde{X}_\varepsilon \varepsilon, \tilde{X}_\varepsilon^\# \varepsilon, \tilde{Q}_\varepsilon, \tilde{Q}_\varepsilon \circ \tilde{X}_\varepsilon).$$

If we prove that $\tilde{\Xi}_\varepsilon$ converges to an enhancement $\tilde{\Xi}$

$$\tilde{\Xi} = (\chi, X, X^\#X, X^\#X^\#, X^\#X, X^\#X^\#X^\#X, Q, Q \circ X),$$

in suitable topologies, where $\chi$ is a constant, we will have automatically that $(v_\varepsilon, v'_\varepsilon, v^Q_\varepsilon) \rightarrow (u, u', u^Q)$ in the appropriate topologies. It is not our aim here to fully develop this sketch of proof, especially because a non-trivial part consists in proving the convergence of the stochastic data for which some powerful machinery has been devised in the paper by Hairer and Quastel [HQ15].
Let us just mention some specific mechanisms at work in the convergence result. The enhanced data for \( v^\varepsilon \) involves non-linear functions of the random field \( \varepsilon^{1/2}X_\varepsilon \). A direct computation shows that for any given \((t, x)\) the family of random variables \( (\varepsilon^{1/2}X_\varepsilon(t, x))_\varepsilon \) converges to a Gaussian random variable with a finite variance. In general the covariance of the random field \( X_\varepsilon \) is given by

\[
Q_\varepsilon(y - z, t - s) = \mathbb{E}[X_\varepsilon(t, y)X_\varepsilon(s, z)] = \varepsilon^{-1}(e^{\Delta (t-s)/\varepsilon^2}C)((y - z)/\varepsilon).
\]

This quantity allows for various bounds:

\[
|Q_\varepsilon(y - z, t - s)| \lesssim \varepsilon^{-1} \wedge (t - s)^{-1/2} \wedge (y - z)^{-1}.
\]

Some Gaussian analysis can be used to show that if we have a sequence of polynomial functions \((F_\varepsilon : \mathbb{R} \to \mathbb{R})_\varepsilon \) such that \( \mathbb{E}[F_\varepsilon(\varepsilon^{1/2}X_\varepsilon)] = 0 \) then for all \( p \geq 1 \) and under some technical assumptions we have for every \( \psi \in C^\infty_c(\mathbb{R}, \mathbb{R}) \)

\[
\psi(t)F_\varepsilon(\varepsilon^{1/2}X_\varepsilon(t, x)) \to 0
\]

almost surely along subsequences in \( B^\infty_\kappa(\mathbb{R} \times \mathbb{T}) \) for some \( \kappa > 0 \). Moreover, if the component of \( F_\varepsilon(\varepsilon^{1/2}X_\varepsilon) \) in the first chaos of the random field \( X_\varepsilon \) vanishes, then the above convergence can be improved at the cost of reducing the space-time regularity:

\[
\varepsilon^{-1/2-\kappa}\psi(t)F_\varepsilon(\varepsilon^{1/2}X_\varepsilon(t, x)) \to 0
\]

as a space–time distribution of parabolic regularity \(-1/2 - \kappa\), almost surely at least along subsequences. See also [GP16] where similar results are derived in a specific stationary setting based on the chaos expansion under the stationary measure of \( X_\varepsilon \).

### 7 Anderson Hamiltonian

Paracontrolled distributions and related tools can not only be used to solve singular SPDEs, they also allow us to construct certain operators that are a priori ill-defined. Consider for example the parabolic Anderson model \( \partial_t u = \Delta u + u\xi \), that is gPAM with \( G(u) = u \). If we consider the Anderson Hamiltonian

\[
\mathcal{H}u = (\Delta + \xi)u,
\]

then formally the solution to the parabolic Anderson model is given by \( u(t) = e^{t\mathcal{H}}u_0 \), where \((e^{t\mathcal{H}})_{t \geq 0}\) is the semigroup generated by \( \mathcal{H} \). So by understanding \( \mathcal{H} \) we should also gain a better understanding of the parabolic equation. In particular it will be interesting to study the spectrum of \( \mathcal{H} \) and the structure of its eigenfunctions in order to learn something about the long time behavior of \( u(t) \).

We would like to see \( \mathcal{H} \) as an unbounded operator on \( L^2(\mathbb{T}^d) \) and not \( C^\infty(\mathbb{T}^d) \), because \( L^2 \) is a Hilbert space and the spectral analysis of operators is much easier on Hilbert spaces than on Banach spaces. For \( d = 1 \) Fukushima and Nakao [FN77] constructed \( \mathcal{H} \) already in 1977, but the case \( d \in \{2, 3\} \) was only very recently understood by Allez and Chouk [AC15]. In the following we sketch their results for \( d = 2 \). Of course, there is no problem to make sense of \( \mathcal{H}u \) if \( u \) is a smooth function. The problem is rather that \( \mathcal{H}u \) should be in \( L^2 \), but for \( u \in C^\infty \) the product \( \xi u \) will not be better behaved than \( \xi \) because the multiplication with a smooth function does not increase the regularity. Since \( \xi \) is a distribution and not an \( L^2 \) function, \( C^\infty \) will not be contained in the domain of \( \mathcal{H} \)!

On the other side if \( u \) is too irregular, the product \( \xi u \) may not be defined. So the idea of Allez and Chouk is to define a domain of paracontrolled functions \( u \) for which \( \mathcal{H}u \) takes
Strictly speaking it is not possible to invert the Laplace operator and we have to shift it and consider $(1 - \Delta)^{-1}$ instead, but for simplicity we ignore this here.

values in $L^2$. For $\alpha \in \mathbb{R}$ we write $H^\alpha = B^\alpha_{2,2}$ for the $L^2$-Sobolev space with regularity $\alpha$ and let

$$D = \{ u \in H^{1-} : u^4 = u - u \langle X \rangle \in H^{2-} \},$$

where $X$ is to be determined. Now we have to deal with Besov spaces other than $B^\infty_{\infty,\infty}$, but it is easy to see that we have analogous estimates for the paraproduct and resonant term on $H^\alpha$ spaces, more precisely

$$\ast \prec \ast : H^\alpha \times C^\beta \rightarrow H^{\beta \wedge (\beta + \alpha)}, \quad \alpha, \beta \in \mathbb{R},$$

$$\ast \ast \prec \ast : H^\alpha \times C^\beta \rightarrow C^{\beta + \alpha}, \quad \alpha + \beta > 0.$$

The commutator estimate also extends to more general Besov spaces, see [PT16]: For $\alpha \in (0, 1)$ and $\beta, \gamma \in \mathbb{R}$ with $\beta + \gamma < 0$ and $\alpha + \beta + \gamma > 0$ we have

$$C : H^\alpha \times C^\beta \times C^\gamma \rightarrow H^{\alpha + \beta + \gamma}.$$

Recall that $\xi \in C^{\gamma - 1}$, so for $X \in C^{\gamma - 1}$ and $u \in D$ we obtain

$$\Delta u + u\xi = (\Delta(u \langle X \rangle - u \langle \Delta X \rangle) + \Delta u^4 + u \langle \Delta X \rangle^4 + u \langle \xi \rangle + u \langle X \xi \rangle + u \xi)_{H^{0-}}$$

$$+ u\xi_{H^{0-}} + C(u, X, \xi) + u \langle X \xi \rangle. \quad (20)$$

Choosing $X$ such that $\Delta X = -\xi$ we get $\mathcal{H}u \in H^{0-}$ for all $u \in D$, and therefore $\mathcal{H}$ is an unbounded operator on $H^{0-}$ with domain $D$. Moreover, $D$ is dense in $H^{0-}$ and a Hilbert space when equipped with the norm $\|u\|_D = \|u\|_{H^{1-}} + \|u^4\|_{H^{2-}}$. But we are interested in the spectral theory for $\mathcal{H}$ and a generic $u \in D$ can never be an eigenfunction because $\mathcal{H}u \in H^{0-} \not\subset D$. To find the eigenfunctions of $\mathcal{H}$ we should identify a subspace of $D$ on which $\mathcal{H}$ has better regularity. The idea of Allez and Chouk is to consider what they call strongly paracontrolled distributions: From the expansion (20) we see that if we want $\mathcal{H}u \in H^{1-}$, then all terms of regularity $H^{0-}$ on the right hand side should cancel up to a remainder of regularity $H^{1-}$, so we should have

$$(\Delta(u \langle X \rangle - u \langle \xi \rangle) + \Delta u^4 + u \langle \xi \rangle + u \langle X \langle \xi \rangle \rangle + u\xi) \in H^{1-},$$

or in other words

$$-\Delta u^4 = (\Delta(u \langle X \rangle - u \langle \xi \rangle) + u \langle \xi \rangle + u \langle X \xi \rangle) + uR,$$

where $uR \in H^{1-}$, from where we get

$$u^4 = (-\Delta)^{-1}(\Delta(u \langle X \rangle - u \langle \xi \rangle) + u \langle \xi \rangle + u \langle X \xi \rangle) + u^{2\xi} = : \Phi(u) + u^{2\xi},$$

where $\Phi(u)$ is defined through the equation and $u^{2\xi} \in H^{3-}$. Thus, the space of strongly paracontrolled distributions is

$$\text{dom}(\mathcal{H}) = \{ u \in H^{1-} : u^{2\xi} = u - u \langle X \rangle - \Phi(u) \in H^{3-} \} \subset D,$$

and for $u \in \text{dom}(\mathcal{H})$ we get

$$\mathcal{H}u = \Delta u + u\xi = (\Phi(u) + u^{2\xi})\xi + C(u, X, \xi) \in H^{1-} \subset L^2.$$
It is not at all trivial that \( \text{dom}(\mathcal{H}) \) contains more functions than just 0, but in [AC15] it is even shown that \( \text{dom}(\mathcal{H}) \) is dense in \( L^2 \). Moreover, it is shown that \( \mathcal{H} \) is a symmetric operator:

\[
\langle \mathcal{H}u, v \rangle_{L^2} = \langle u, \mathcal{H}v \rangle_{L^2}
\]

for all \( u, v \in \text{dom}(\mathcal{H}) \), and there exists a decreasing sequence \( \lambda_1 \geq \lambda_2 \geq \ldots \) of real eigenvalues with \( \lim_{n \to \infty} \lambda_n = -\infty \) and an \( L^2 \)-orthonormal basis of corresponding eigenvectors \((e_n)_{n \in \mathbb{N}}\) such that \( \mathcal{H}e_n = \lambda_n e_n \) for all \( n \in \mathbb{N} \) and such that

\[
\mathcal{H}u = \sum_{n=1}^{\infty} \lambda_n \langle u, e_n \rangle_{L^2}, \quad u \in \text{dom}(\mathcal{H}).
\]

There is just one thing that we omitted: as for the parabolic Anderson model it is of course necessary to renormalise the operator, because if \( (\xi_t)_{t \geq 0} \) is a convolution approximation of \( \xi \) and \( -\Delta X_t = \xi_t \), then the term \( X_t \circ \xi_t \) does not converge but only \( X_t \circ \xi_t - c_t \) converges for a suitable sequence of diverging constants \((c_t)_{t \geq 0}\). Replacing \( X_t \circ \xi_t \) by \( X_t \circ \xi_t - \infty \) has the effect of changing \( \mathcal{H}u = \Delta u + u\xi \) to

\[
\Delta u + u(\xi - \infty) = \Delta u + u\xi + u\xi + C(u, X, \xi) + u(X \circ \xi - \infty).
\]

From here we see that if \( \xi \) is the space white noise, the operator \( \mathcal{H} \) cannot be continuously extended from \( \text{dom}(\mathcal{H}) \) to the smooth functions because for \( u \in C^\infty \) the product \( u\xi \) does not create any divergences so \( u(\xi - \infty) \) does not make any sense!

Allez and Chouk [AC15] then proceed to study how the largest eigenvalue \( \lambda_1 \) behaves in the white noise case, and they show that there exist constants \( C_1, C_2 > 0 \) such that

\[
e^{-C_1 x} \leq \mathbb{P}(\lambda_1 \geq x) \leq e^{-C_2 x} \quad (21)
\]

for \( x \to \infty \). From here we learn at least heuristically that at large times the solution \( u \) to the parabolic Anderson model should not have any moments, because

\[
u(t) = e^{t\mathcal{H}}u_0 = \sum_{n=1}^{\infty} e^{\lambda_n t} \langle u_0, e_n \rangle_{L^2},
\]

and omitting the contribution from all eigenvalues except \( \lambda_1 \) we get

\[
\mathbb{E}[|e^{t\lambda_1} \langle u_0, e_n \rangle_{L^2}|^p] = \mathbb{E}[e^{t\lambda_1} |\langle u_0, e_n \rangle_{L^2}|^p].
\]

By (21) we have \( \mathbb{E}[e^{t\lambda_1}] = \infty \) as soon as \( tp > C_1 \), so in that case we expect that also \( \mathbb{E}[|u(t)|^p] = \infty \). But it remains an open problem how to make this intuitive argumentation rigorous.

## 8 Singular martingale problem

Similar ideas that we developed to study the Anderson Hamiltonian \( \mathcal{H}u = (\Delta + \xi)u \) have also been used by Cannizzaro and Chouk [CC15], inspired by [DD16], to make sense of certain diffusions with distributional drift: Let \( \xi \in C^{\alpha-1}(\mathbb{R}) \) with \( \alpha \in (1/3, 1/2) \) and consider the SDE \( x: \mathbb{R}_+ \to \mathbb{R} \),

\[
dx_t = \xi(x_t)dt + \sqrt{2}dw_t,
\]

where \( w \) is a Brownian motion. Of course, this equation does not make any sense because \( \xi \) is a distribution and cannot be evaluated in \( x_t \). We still formally write down the

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\( ^{\#} \)Besov spaces on \( \mathbb{R} \) are defined exactly in the same way as on \( T \) and they have essentially the same properties.
martingale problem and call a continuous stochastic process $x$ a solution if for all suitable $u$ the process

$$ M_t^u = u(x_t) - u(x_0) - \int_0^t \mathcal{G}u(x_s) \, ds, \quad t \geq 0, $$

is a continuous martingale, where $\mathcal{G}u = \xi \partial_x u + \Delta u$. But just as for the Anderson Hamiltonian the problem is that $\mathcal{G}u$ is only a distribution whenever $u$ is a smooth function, and therefore we first need to identify a suitable domain of functions for which $\mathcal{G}u$ is continuous. We can do this by solving the equation

$$ (\mathcal{G} - \lambda)u = \varphi, \quad (22) $$

for $\varphi \in C^b(\mathbb{R})$, the continuous and bounded functions on $\mathbb{R}$, and $\lambda > 0$. Then we get for the solution $u$ that $\mathcal{G}u = \varphi + \lambda u$, so provided that $u$ itself is a continuous function the martingale problem above makes sense. To solve (22) we make the paracontrolled Ansatz $u = u' \prec X + u^\sharp$ with $u' \in C^\alpha(\mathbb{R})$, $(\Delta - \lambda)X = \partial_x \xi \in C^\alpha(\mathbb{R})$, and $u^\sharp \in C^{2\alpha}(\mathbb{R})$, and we reformulate the equation as

$$ (\Delta - \lambda)u = \xi \partial_x u + \varphi. $$

From here it is not difficult to see that for $\lambda$ large enough (depending only on $\xi$ and $X \circ \xi$ but not on $\varphi$) there exists a unique paracontrolled solution $u$ to the equation, and the space of paracontrolled functions $u$ which solve $(\mathcal{G} - \lambda)u = \varphi$ for some $\varphi \in C^b(\mathbb{R})$ is a domain for $\mathcal{G}$. Moreover, there exists a unique (in law) solution $x$ to the martingale problem defined above. For details see [CC15].

References

[AC15] Romain Allez and Khalil Chouk. The continuous Anderson Hamiltonian in dimension two. arXiv preprint arXiv:1511.02718, 2015.

[AR91] Sergio Albeverio and Michael Röckner. Stochastic differential equations in infinite dimensions: solutions via Dirichlet forms. Probability theory and related fields, 89(3):347–386, 1991.

[BB16a] Ismaël Bailleul and Frederic Bernicot. Heat semigroup and singular PDEs. Journal of Functional Analysis, 270(9):3344–3452, 2016.

[BB16b] Ismaël Bailleul and Frédéric Bernicot. Higher order paracontrolled calculus. arXiv preprint arXiv:1609.06966, 2016.

[BCD11] Hajer Bahouri, Jean-Yves Chemin, and Raphal Danchin. Fourier Analysis and Nonlinear Partial Differential Equations. Springer, January 2011.

[BDH16] Ismael Bailleul, Arnaud Debussche, and Martina Hofmanova. Quasilinear generalized parabolic Anderson model. arXiv preprint arXiv:1610.06726, 2016.

[BHZ16] Yvain Bruned, Martin Hairer, and Lorenzo Zambotti. Algebraic renormalisation of regularity structures. arXiv:1610.08468 [math], October 2016. arXiv: 1610.08468.

[Bon81] Jean-Michel Bony. Calcul symbolique et propagation des singularités pour les équations aux dérivées partielles non linéaires. In Annales scientifiques de l’École Normale supérieure, volume 14, pages 209–246, 1981.

[CC13] Rémi Catellier and Khalil Chouk. Paracontrolled distributions and the 3-dimensional stochastic quantization equation. arXiv preprint arXiv:1310.6869, 2013.

[CC15] Giuseppe Cannizzaro and Khalil Chouk. Multidimensional SDEs with singular drift and universal construction of the polymer measure with white noise potential. arXiv preprint arXiv:1501.04751, 2015.
[CF14] Khalil Chouk and Peter K Friz. Support theorem for a singular semilinear stochastic partial differential equation. arXiv preprint arXiv:1409.4250, 2014.

[CFG17] Giuseppe Cannizzaro, Peter K Friz, and Paul Gassiat. Malliavin calculus for regularity structures: The case of gPAM. Journal of Functional Analysis, 272(1):363–419, 2017.

[CGP16] Khalil Chouk, Jan Gairing, and Nicolas Perkowski. An invariance principle for the two-dimensional parabolic Anderson model with small potential. arXiv preprint arXiv:1609.02471, 2016.

[CH16] Ajay Chandra and Martin Hairer. An analytic BPHZ theorem for regularity structures. arXiv:1612.08138 [math-ph], December 2016. arXiv: 1612.08138.

[DD16] François Delarue and Roland Diel. Rough paths and 1d SDE with a time dependent distributional drift: application to polymers. Probability Theory and Related Fields, 165(1-2):1–63, 2016.

[DPD03] Giuseppe Da Prato and Arnaults Debussche. Strong solutions to the stochastic quantization equations. The Annals of Probability, 31(4):1900–1916, 2003.

[FH14] Peter K. Friz and Martin Hairer. A Course on Rough Paths: With an Introduction to Regularity Structures. Springer, August 2014.

[FH16] Tadahisa Funaki and Masato Hoshino. A coupled KPZ equation, its two types of approximations and existence of global solutions. arXiv preprint arXiv:1611.00498, 2016.

[FN77] Masatoshi Fukushima and Shintaro Nakao. On spectra of the Schrödinger operator with a white Gaussian noise potential. Zeitschrift für Wahrscheinlichkeitstheorie und Verwandte Gebiete, 37(3):267–274, 1977.

[GIP15] Massimiliano Gubinelli, Peter Imkeller, and Nicolas Perkowski. Paracontrolled distributions and singular PDEs. Forum of Mathematics. Pi, 3:e6, 75, 2015.

[GP15] M. Gubinelli and N. Perkowski. Lectures on singular stochastic PDEs. Ensaios Matemáticos, 29, 2015. arXiv: 1502.00157.

[GP16] Massimiliano Gubinelli and Nicolas Perkowski. The Hairer–Quastel universality result in equilibrium. arXiv:1602.02428 [math-ph], February 2016. arXiv: 1602.02428.

[GP17] Massimiliano Gubinelli and Nicolas Perkowski. KPZ reloaded. Communications in Mathematical Physics, 349(1):165–269, 2017.

[Gub04] M. Gubinelli. Controlling rough paths. Journal of Functional Analysis, 216(1):86–140, 2004.

[Hai11] M. Hairer. Rough stochastic PDEs. Communications on Pure and Applied Mathematics, 64(11):1547–1585, 2011.

[Hai14] M. Hairer. A theory of regularity structures. Inventiones mathematicae, 198(2):269–504, March 2014.

[HQ15] Martin Hairer and Jeremy Quastel. A class of growth models rescaling to KPZ. arXiv:1512.07845 [math-ph], December 2015. arXiv: 1512.07845.

[Kup16] Antti Kupiainen. Renormalization Group and Stochastic PDEs. Annales Henri Poincaré. A Journal of Theoretical and Mathematical Physics, 17(3):497–535, 2016.

[LCL07] Terry J. Lyons, Michael J. Caruana, and Thierry Lévêque. Differential Equations Driven by Rough Paths: Ecole d’Eté de Probabilités de Saint-Flour XXXIV-2004. Springer, 1 edition, June 2007.
[LQ02] Terry Lyons and Zhongmin Qian. System Control and Rough Paths. Oxford University Press, 2002.

[Lyo98] Terry Lyons. Differential equations driven by rough signals. Revista Matemática Iberoamericana, pages 215–310, 1998.

[Mey81] Yves Meyer. Remarques sur un thôrme de J.-M. Bony. In Rendiconti del Circolo Matematico di Palermo. Serie II, pages 1–20, 1981.

[MW15] Jean-Christophe Mourrat and Hendrik Weber. Global well-posedness of the dynamic $\phi^4$ model in the plane. arXiv preprint arXiv:1501.06191, 2015.

[MW16] Jean-Christophe Mourrat and Hendrik Weber. Global well-posedness of the dynamic $\phi^3_3$ model on the torus. arXiv preprint arXiv:1601.01234, 2016.

[MWX16] Jean-Christophe Mourrat, Hendrik Weber, and Weijun Xu. Construction of $\phi^4_3$ diagrams for pedestrians. arXiv preprint arXiv:1610.08897, 2016.

[OW16] Felix Otto and Hendrik Weber. Quasilinear SPDEs via rough paths. arXiv preprint arXiv:1605.09744, 2016.

[PT16] David J Prömel and Mathias Trabs. Rough differential equations driven by signals in Besov spaces. Journal of Differential Equations, 260(6):5202–5249, 2016.

[RZZ15] Michael Röckner, Rongchan Zhu, and Xiangchan Zhu. Restricted Markov uniqueness for the stochastic quantization of $P(\Phi)_2$ and its applications. arXiv preprint arXiv:1511.08030, 2015.

[ZZ14] Rongchan Zhu and Xiangchan Zhu. Approximating three-dimensional Navier-Stokes equations driven by space-time white noise. arXiv preprint arXiv:1409.4864, 2014.

[ZZ15] Rongchan Zhu and Xiangchan Zhu. A Wong-Zakai theorem for $\phi^4_3$ model. arXiv preprint arXiv:1504.04143, 2015.