Circulation and Energy Theorem Preserving Stochastic Fluids

Theodore D. Drivas
Department of Mathematics, Princeton University, Princeton, NJ 08544, United States (tdrivas@math.princeton.edu)

Darryl D. Holm
Department of Mathematics, Imperial College, London SW7 2AZ, UK (d.holm@imperial.ac.uk)

(MS Received 10 September 2018; accepted 18 June 2019)

Smooth solutions of the incompressible Euler equations are characterized by the property that circulation around material loops is conserved. This is the Kelvin theorem. Likewise, smooth solutions of Navier–Stokes are characterized by a generalized Kelvin’s theorem, introduced by Constantin–Iyer (2008). In this note, we introduce a class of stochastic fluid equations, whose smooth solutions are characterized by natural extensions of the Kelvin theorems of their deterministic counterparts, which hold along certain noisy flows. These equations are called the stochastic Euler–Poincaré and stochastic Navier–Stokes–Poincaré equations respectively. The stochastic Euler–Poincaré equations were previously derived from a stochastic variational principle by Holm (2015), which we briefly review. Solutions of these equations do not obey pathwise energy conservation/dissipation in general. In contrast, we also discuss a class of stochastic fluid models, solutions of which possess energy theorems but do not, in general, preserve circulation theorems.

Keywords: stochastic fluid equations; variational principle; Kelvin theorem

1. Introduction

In 1869, Lord Kelvin (Sir William Thomson) [46] discovered a beautiful property of smooth solutions of the incompressible Euler equations. Namely, the circulation of velocity around any closed loop advected by an ideal fluid is conserved. More precisely, let the spatial domain of flow be $\Omega = \mathbb{T}^d$ or $\mathbb{R}^d$, and suppose the fluid velocity $u_t := u(x, t) : \Omega \times [0, T] \to \mathbb{R}^d$ solves the incompressible Euler equations,

$$\begin{align*}
\partial_t u_t + (u_t \cdot \nabla) u_t &= -\nabla p_t, \\
\nabla \cdot u_t &= 0,
\end{align*}$$

with scalar pressure function $p_t$, determined by solving the Poisson equation

$$-\Delta p_t = (\nabla \otimes \nabla) : (u_t \otimes u_t),$$

which enforces incompressibility at each time, $t$. The Kelvin theorem states that any smooth Euler solution $u_t$ has the property that for all loops $\Gamma \subset \Omega$, the circulation
integral satisfies,
\[ \oint_{X_t(\Gamma)} u_t \cdot d\ell = \oint_{\Gamma} u_0 \cdot d\ell, \quad (1.3) \]
where \( X_t \) is Lagrangian flow satisfying \( \dot{X}_t = u_t(X_t) \), \( X_0 = \text{id} \). The Kelvin theorem offers an elegant interpretation of the Lagrangian laws of vortex motion written down by Helmholtz in 1858 [36].

Remarkably, the converse implication also holds. That is, any sufficiently regular incompressible velocity field possessing the property (1.3) for all times \( t \in [0,T] \) and for any closed, rectifiable loop \( \Gamma \subset \Omega \) must, in fact, be a smooth Euler solution. This follows readily from (1.3), since its time derivative implies that
\[ \oint_{X_t(\Gamma)} (\partial_t u_t + (u_t \cdot \nabla) u_t + (\nabla u_t)^T \cdot u_t) \cdot d\ell = 0 \quad (1.4) \]
for all loops \( \Gamma \) and all times \( t \in [0,T] \). In particular, equation (1.4) holds for the rectifiable loop \( \Gamma = X_t^{-1}(\Gamma') \) for any fixed \( t' \in [0,T] \) (since \( X_t \) is a diffeomorphism). For such a loop, evaluating Eq. (1.4) at time \( t = t' \) shows that the line integral vanishes when \( X_t(\Gamma) \) above is replaced by an arbitrary loop \( \Gamma' \). From Stokes theorem, equation (1.4) holds for all loops \( \Gamma' \subset \Omega \), if and only if there exists a scalar function \( \pi_t = \pi(x,t) \) such that the integrand is equal to the gradient of this potential \( \nabla \pi_t \) for all \( (x,t) \in \Omega \times [0,T] \). Then, using the identity \( (\nabla u_t)^T \cdot u_t = \nabla |u_t|^2/2 \), one finds that equation (1.1) holds with \( p_t \) replaced by \( q_t := 1/2 |u_t|^2 - \pi_t \). Finally, to enforce incompressibility of \( u_t \), the scalar function \( q_t \) solves equation (1.2) thus fixing it as the pressure \( q_t = p_t \) (up to a constant). Therefore, one may say that smooth solutions of the Euler equations in the domain \( \Omega \) are characterized by the Kelvin theorem. It is worth noting that this equivalence was already realized by Lord Kelvin in his original 1869 paper [46].

Kelvin’s theorem has long been recognized as centrally important to the understanding of deterministic, smooth, ideal fluid dynamics. Its geometric meaning is discussed in Appendix A. In the present work, we take a geometric and variational approach to answering the following question: How would Kelvin’s theorem be changed, if the fluid flow were stochastic?

Summary of the present work. The present work follows Holm [37] in which a family of stochastic partial differential equations (SPDEs) for fluid dynamics was derived from a stochastic variational principle as Euler–Poincaré equations. We will prove that these stochastic Euler–Poincaré equations are, in fact, characterized by the pathwise Kelvin theorem (1.3) in which the Lagrangian flow \( X_t \) is a particular stochastic process. The stochastic variational principle from which these stochastic Euler–Poincaré equations arose is reviewed in Appendix B. A key feature of their derivation is the decomposition of the Lagrangian flow map into fast and slow components. The fast motion of the Lagrangian trajectory is represented by a stochastic process, whose correlate statistics are to be calibrated from data as in [12, 13]. The stochastic decomposition proposed in [37] was later derived using multi-time homogenization by Cotter et al. [11]. The well-posedness of the stochastic Euler fluid version of these stochastic Euler–Poincaré equations in three
dimensions was established in [14] for initial conditions in an appropriate Sobolev space. This well-posedness result is discussed more specifically in remark 2.4 below.

**Stochastic Euler–Poincaré equations.** Next we introduce the stochastic Euler fluid equations in [37], whose Kelvin circulation properties will be investigated in the remainder of the paper. The derivation of these stochastic fluid equations in standard geometric variational notation is reviewed in the Appendices.

Let \((\Xi,\mathcal{F},P)\) be a complete filtered probability space with a filtration \(\mathcal{F}_t\) of right continuous \(\sigma\)-algebras \((\mathcal{F}_t)_{t \geq 0}\). Let \(\{W_t^{(k)}\}_{k \in \mathbb{N}}\) be a collection of \(\mathcal{F}_t\)-adapted independent 1-dimensional Brownian motions in \(\mathbb{R}\). The Stratonovich form of the circulation-theorem preserving stochastic Euler–Poincaré equations introduced in [37] for the 1-form \(u_t\) reads

\[
du_t + \mathbb{P}(\mathcal{L}^T u_t)dt + \sum_k \mathbb{P}(\mathcal{L}^T u_t) \circ dW_t^{(k)} = 0, \quad \text{with} \quad u_t|_{t=0} = u_0. \tag{1.5}
\]

Here, \(\mathbb{P}\) is the dual for 1-forms of the standard Leray projection operator for vector fields, as discussed in Appendix A. The symbol \(\circ\) denotes the Stratonovich sense of the stochastic product, the collection \(\{\xi^{(k)}\}_{k \in \mathbb{N}}\) contains fixed, deterministic divergence-free vector fields \(\xi^{(k)} : \Omega \rightarrow \mathbb{R}^d\), to be calibrated in practice from data, e.g., as in [12, 13], and we define the operator \(\mathcal{L}^T\) as

\[
\mathcal{L}^T \xi u_t := v \cdot \nabla u_t + (\nabla v)^T \cdot u_t \\
= -v \times \text{curl} u_t + \nabla (v \cdot u_t) \quad \text{in 3D}. 
\tag{1.6}
\]

In index notation, \((\mathcal{L}^T u)_i = v^j \partial_j u_i + (\partial_i v^j)u_j\). The geometric justification for choosing to write the nonlinearity in the \(\mathcal{L}^T\) form is explained in Appendix A. Smooth solutions of the stochastic Euler–Poincaré equations in Eqns. (1.5) possess a stochastic Kelvin theorem, which we describe in theorem 1.3 below.

**2. Main Results**

In this paper, we prove that the stochastic Euler–Poincaré equations are, in fact, characterized by the pathwise Kelvin theorem (1.3). For this purpose, we consider a class of abstract stochastic Itô SPDEs

\[
du_t + \mathbb{P}f_t dt + \sum_k \mathbb{P}\sigma_t^{(k)} dW_t^{(k)} = 0, \quad \text{with} \quad u_t|_{t=0} = u_0. \tag{2.1}
\]

The system (2.1) maintains incompressibility \(\nabla \cdot u_t = 0\) while its solutions exist. Equation (2.1) for the 1-form \(u_t\) is to be understood in the weak sense: for any solenoidal test vector field \(\varphi \in C_0^\infty(\Omega)\), the following equality holds

\[
\langle u_t, \varphi \rangle_{L^2} = \langle u_0, \varphi \rangle_{L^2} - \int_0^t \langle f_t, \varphi \rangle_{L^2} dt - \sum_k \int_0^t \langle \sigma_t^{(k)}, \varphi \rangle_{L^2} dW_t^{(k)}. \tag{2.2}
\]

We prove the following result.
Hence, the stochastic Euler–Poincaré equations (2.7) as the unique equations for which (2.3).

Then, $u_t$ is a smooth solution of (2.1) on $[0, T] \times \Omega$ with

$$f_t = L_{u_t}^T u_t - \sum_k \frac{1}{2} \ell_{\xi(k)}^T (L_{\xi(k)}^T u_t),$$

(2.4)

and

$$\sigma_t^{(k)} = L_{\xi(k)}^T u_t,$$

(2.5)

if and only if, for every rectifiable loop $\Gamma \subset \Omega$, $u_t$ has the property that for all $t \in [0, T]$,

$$\oint_{X_t(\Gamma)} u_t \cdot d\ell = \oint_{\Gamma} u_0 \cdot d\ell, \quad \mathbb{P} \text{ a.s.}$$

(2.6)

Remark 2.2 (Itô form of the Stochastic Euler–Poincaré Equations). The Itô form of equation (1.5) reads

$$du_t + \mathbb{P} \left( L_{u_t}^T u_t - \frac{1}{2} \sum_k L_{\xi(k)}^T (L_{\xi(k)}^T u_t) \right) dt + \sum_k \mathbb{P} L_{\xi(k)}^T u_t dW_t^{(k)} = 0.$$ (2.7)

This follows from the Stratonovich-to-Itô conversion

$$\int_0^t \mathbb{P} L_{\xi(k)}^T u_s \circ dW_s^{(k)} = \int_0^t \mathbb{E} L_{\xi(k)}^T u_s dW_s^{(k)} + \frac{1}{2} \mathbb{E} \left[ L_{\xi(k)}^T u_t, W^{(k)} \right],$$

where $[\cdot, \cdot]_t$ denotes the quadratic variation and where we have used equation (1.5) to compute this cross-variation. To obtain (2.7), we note that for any 1-form $v$, we have $\mathbb{P} L_{\xi}^T \mathbb{P} v = \mathbb{P} L_{\xi}^T v$. To see this, write $\mathbb{P} v = v + \nabla q$ for some scalar $q$ and note that $\ell_{\xi}^T \nabla q$ is a gradient,

$$L_{\xi}^T \nabla q = (\xi \cdot \nabla) \nabla q + \nabla \xi \cdot \nabla q = \nabla (\xi \cdot \nabla q).$$ (2.8)

Hence, $\mathbb{P} L_{\xi}^T \nabla q = 0$ and equation (2.7) follows. In view of theorem 2.1, we recover the stochastic Euler–Poincaré equations (2.7) as the unique equations for which smooth solutions obey pathwise circulation conservation along the stochastic flow (2.3).

Remark 2.3 (Regularity of Flow). Provided that $\sum_k \|\xi(k)\|_{C^{n+3, \alpha'}(\Omega)}^2 < \infty$ for some $n \in \mathbb{N}$ and $\alpha' \in (0, 1)$, and $u \in C(0, T; C^{n+1, \alpha}(\Omega))$ for any $\alpha \in (0, \alpha')$, then equation (2.3) generates a flow of $C^{n+1, \alpha}$–diffeomorphisms of $\Omega$ [51, 52]. Moreover,
The inverse map $A_t := X_t^{-1}$ exists and belongs to the same space $C(0, T; C^{n+1,\alpha}(\Omega))$ and the gradient belongs to $\nabla X_t \in C(0, T; C^{n,\alpha}(\Omega))$. For $n \geq 2$, this sufficient regularity to justify the computations of the present paper, in particular the use of the Itô–Wentzell formula [50, 51].

**Remark 2.4 (Local Existence and Regularity for Euler–Poincaré Fluids).** Well-posedness for equations (2.1) with (2.4) and (2.5) has recently been established in [14]. In §3.3 of [14], it is shown there exists (for data in the appropriate Sobolev space) a maximal stopping time $\tau_{\text{max}} : \Xi \mapsto [0, \infty)$ and a unique solution $u \in C(0, \tau; W^{3,2}(\mathbb{T}^3; \mathbb{R}^3))$ for all $\tau \leq \tau_{\text{max}}$. Subsequently, [25] established local existence of (1.5)–(1.6) in Hölder spaces $C(0, T; C^{n+1,\alpha}(\Omega))$ for some $n \in \mathbb{N}$ and some $\alpha \in (0, 1)$, by using the Weber formula (2.17) and following the Eulerian–Lagrangian scheme of Constantin [7]. Thus, in view of remark 2.3, regularity in the appropriate Hölder spaces can be taken as the precise meaning of ‘smooth’ in theorem 2.1 as well as in proposition 2.5 and theorem 2.12 appearing below.

The key to the proof of theorem 2.1 is a general formula for the transport of circulations along the stochastic flow (2.3), where the velocity $u_t$ is a stochastic process driven by the same Brownian noise $\{W_t^{(k)}\}_{k \in \mathbb{N}}$.

**Proposition 2.5 (Stratonovich Stochastic Circulation Transport).** Fix smooth vector fields $\xi^{(k)} : \Omega \mapsto \mathbb{R}^d$. Let $u_t : [0, T] \times \Omega \mapsto \mathbb{R}^d$ be a smooth solution of equation (2.1) and $X_t := X_t(x)$ be the stochastic flow defined by the SDE

$$dX_t(x) = u_t(X_t(x))dt + \sum_k \xi^{(k)}(X_t(x)) \circ dW_t^{(k)}, \quad X_0(x) = x. \tag{2.9}$$

Then, for any rectifiable loop $\Gamma \subset \Omega$, the following holds for $t \in [0, T]$

$$d\int_{X_t(\Gamma)} u_t \cdot d\ell = \int_{X_t(\Gamma)} \left( \mathcal{L}^T_{u_t} u_t - f_t \right) dt \cdot d\ell + \sum_k \int_{X_t(\Gamma)} \left( \frac{1}{2} \mathcal{L}^T_{\xi^{(k)}} \mathcal{L}^T_{\xi^{(k)}} u_t - \mathcal{L}^T_{\xi^{(k)}} \sigma_t^{(k)} \right) dt \cdot d\ell + \sum_k \int_{X_t(\Gamma)} \left( \mathcal{L}^T_{\xi^{(k)}} u_t - \sigma_t^{(k)} \right) dW_t^{(k)} \cdot d\ell. \tag{2.10}$$

**Remark 2.6 (Itô–Wentzell formula).** The noise appearing in the flow (2.9) is the same noise that drives the stochastic evolution of $u_t$. Consequently, these objects are correlated and to compute the rate of change of circulation we employ the Itô–Wentzell formula. This formula results in the presence of the term $\mathcal{L}^T_{\xi^{(k)}} \sigma_t^{(k)}$ in the second line of equation (2.10).

**Remark 2.7 (Pathwise Kelvin theorem along Itô flow).** In contrast with the flow (2.9) in proposition 2.5, we may consider loops which are transported by stochastic
Circulation and Energy Theorem Preserving Stochastic Fluids

2781

flow with Itô noise instead of Stratonovich noise; namely,

\[ dY_t(x) = u_t(Y_t(x))dt + \sum_k \xi^{(k)}(Y_t(x))dW_t^{(k)}, \quad Y_0(x) = x. \]  

(2.11)

These trajectories can be realized as solving an equivalent Stratonovich equation

\[ dY_t(x) = u_t(Y_t(x))dt + b_t(Y_t(x))dt + \sum_k \xi^{(k)}(Y_t(x)) \circ dW_t^{(k)}, \quad Y_0(x) = x. \]  

(2.12)

with vector field

\[ b_t(x) := -\frac{1}{2} \sum_k (\xi^{(k)}(x) \cdot \nabla)\xi^{(k)}(x), \]

known as the noise-induced drift.

Upon treating \( b_t \) as an arbitrary vector field for now, an argument similar to that made to prove proposition 2.5 shows that for any rectifiable loop \( \Gamma \), the following holds

\[
d \oint_{Y_t(\Gamma)} u_t \cdot d\ell = \oint_{Y_t(\Gamma)} (\mathcal{L}^T_{u_t} u_t + \mathcal{L}^T_{b_t} u_t - f_t) \cdot d\ell \\
+ \sum_k \oint_{Y_t(\Gamma)} \left( \frac{1}{2} \mathcal{L}^T_{\xi^{(k)}} (\mathcal{L}^T_{\xi^{(k)}} u_t) - \mathcal{L}^T_{\xi^{(k)}} \sigma_t^{(k)} \right) \cdot d\ell \\
+ \sum_k \oint_{Y_t(\Gamma)} \left( \mathcal{L}^T_{\xi^{(k)}} u_t - \sigma_t^{(k)} \right) dW_t^{(k)} \cdot d\ell,
\]  

(2.13)

generalizing the formula (2.9) from proposition 2.5. This calculation has proved the following proposition.

**Proposition 2.8 (Itô Stochastic Circulation Transport).** The pathwise Kelvin theorem for a loop advected by the Itô flow defined by (2.11)

\[
\oint_{Y_t(\Gamma)} u_t \cdot d\ell = \oint_{\Gamma} u_0 \cdot d\ell, \quad \mathbf{P} \text{ a.s.}
\]  

(2.14)

characterizes solenoidal vector fields \( u_t \) as smooth solutions of the SPDE (2.1) with

\[
f_t = \mathcal{L}^T_{u_t} u_t - \sum_k \left( \frac{1}{2} \mathcal{L}^T_{\xi^{(k)}} (\mathcal{L}^T_{\xi^{(k)}} u_t) + \mathcal{L}^T_{1/2(\xi^{(k)}.\nabla)\xi^{(k)}} u_t \right),
\]  

(2.15)

\[
\sigma_t^{(k)} = \mathcal{L}^T_{\xi^{(k)}} u_t.
\]  

(2.16)

Upon comparing the drift \( f_t \) in (2.15) with that in (2.4), we see that advection of the loop by the Itô flow \( Y_t \) in (2.11) induces the same double Lie derivative diffusion term as for advection by the Stratonovich flow \( X_t \) in (2.9), plus an additional advective term. Thus, a modification of the drift occurs when the Itô flow (2.11) advects the loop, instead of the Stratonovich flow (2.9). See §1 of [37] for a discussion of a similar issue; quantification of the failure of circulation conservation along loops which are advected by Itô flow, when the velocity satisfies an SPDE whose smooth solutions conserve circulation along loops evolving according to the flow with Stratonovich noise.
Remark 2.9 (Pathwise Weber Formula). A simple consequence of the calculations used in proofs of theorem 2.1 and proposition 2.5 is that smooth solutions of the stochastic Euler–Poincaré equations (1.5)–(1.6) satisfy a pathwise Weber formula:

\[ u_t^\sharp(x) = \mathbb{P}[(\nabla A_t(x))^T u_0(A_t(x))]^\sharp, \quad \mathbb{P}\text{ a.s.} \quad (2.17) \]

where \( A_t = X_t^{-1} \) is the ‘back-to-labels’ map and \( X_t \) is the stochastic flow defined by (2.3) and, when taking projection \( \mathbb{P} \), there is an implied transformation from 1-forms to vector fields by the operation \( \sharp \). See Appendix A for more explanation of the notation \( \sharp \). For a proof of the representation (2.17), see [25]. This result can be expressed also at the level of vorticity \( \omega_t = \text{curl}(u_t) \), where one has an exact Cauchy formula of the form

\[ \omega_t(x) = ((\nabla X_t)\omega_0) \circ A_t(x), \quad \mathbb{P}\text{ a.s.} \quad (2.18) \]

Cauchy’s vorticity representation in (2.18) elucidates what is already apparent directly from (2.6). Namely, the circulation theorem may be expressed, using Stokes theorem, in terms of the flux of vorticity through advected areas. Specifically, letting \( S \) be any smooth bounding surface of the closed loop \( \Gamma \) with \( \Gamma = \partial S \), we have

\[ \oint_{X_t(S)} \omega_t \cdot dS = \oint_S \omega_0 \cdot dS, \quad \mathbb{P}\text{ a.s.} \quad (2.19) \]

Remark 2.10 (Pathwise Energy Preserving Stochastic Fluids). In general for non-constant \( \{\xi^{(k)}_{(k)}\}_{k \in \mathbb{N}} \), the Equations (1.5)–(1.6) do not conserve energy, neither pathwise, nor in expectation. See remark 2.14 for more details. Here, we briefly consider a class of stochastic fluid equations that, by design, do conserve energy pathwise. These can be expressed with Stratonovich noise in terms of the operator \( B(w,v) := \mathbb{P}(w \cdot \nabla v) \)

\[ du_t + B(u_t, dt) + \sum_k \xi^{(k)} \circ dW_t^{(k)}, u_t. \quad (2.20) \]

In Itô form, equation (2.20) reads as equation (2.1) with

\[ f_t := u_t \cdot \nabla u_t - \sum_k \xi^{(k)} \cdot \nabla \mathbb{P}(\xi^{(k)} \cdot \nabla u_t), \quad (2.21) \]

\[ \sigma_t^{(k)} := \xi^{(k)} \cdot \nabla u_t. \quad (2.22) \]

Versions of this model were previously considered in e.g. [24, 56] and discussed in § 5.4 of [22]. We now verify pathwise energy conservation of the model (2.1) with (2.21) and (2.22). This property is most easily and directly established by using Stratonovich calculus and making use of well known properties of the \( B(w,v) \)
Circulation and Energy Theorem Preserving Stochastic Fluids

Since \( \{\xi^{(k)}\}_{k \in \mathbb{N}} \) are divergence-free, we simply have
\[
\frac{1}{2} d\|u_t\|_{L^2}^2 = -(u_t, B(u_t dt + \sum_k \xi^{(k)} \circ dW^{(k)}_t, u_t))_{L^2(\Omega)} = 0. \tag{2.23}
\]

On the other hand, by proposition 2.5 with \( \sigma_t := \sum_k \xi^{(k)} \cdot \nabla u_t \)
and \( f_t := u_t \cdot \nabla u_t \), we have
\[
d\int_{X_t(\Gamma)} u_t \cdot d\ell = \sum_k \int_{X_t(\Gamma)} \left( \frac{1}{2} \mathcal{L}^T_{\xi^{(k)}} (\nabla \xi^{(k)} \cdot u_t) \right) \cdot d\ell \ dt + \sum_k \int_{X_t(\Gamma)} (\nabla \xi^{(k)} \cdot u_t) \cdot d\ell \circ dW^{(k)}_t = \sum_k \int_{X_t(\Gamma)} (\nabla \xi^{(k)} \cdot u_t) \cdot d\ell \circ dW^{(k)}_t.
\]

Thus, unless \( \{\xi^{(k)}\}_{k \in \mathbb{N}} \) are spatially constant, the class of stochastic equations (2.1) with (2.21) and (2.22) which conserve energy pathwise are different with those that possess a pathwise Kelvin circulation theorem.

**Remark 2.11.** Spatially constant noise coefficients \( \{\xi^{(k)}\}_{k \in \mathbb{N}} \) define a privileged class of equations, solutions of which possess both circulation and energy conservation. In particular, when the \( \{\xi^{(k)}\}_{k \in \mathbb{N}} \) are constants, the stochastic Euler–Poincaré equations are essentially deterministic Euler equations in disguise. Specifically, let \( u_t \) solve equation (1.5) and define
\[
v_t(x) := u_t \left( x + \sum_k \xi^{(k)} W^{(k)}_t \right). \tag{2.24}
\]

Then the process \( v_t \) is incompressible \( \nabla \cdot v_t = 0 \) and solves
\[
\partial_t v_t + v_t \cdot \nabla v_t = -\nabla p_t, \tag{2.25}
\]
where \( p_t \) solves the Poisson problem (1.2) to enforce incompressibility of the field \( v_t \).

Thus, formally, \( v_t \) satisfies the usual deterministic Euler equation showing that these two equations have the same form. To see this, suppose that a strong stochastic solution \( u_t \) exists on \( \Omega \times [0, T] \) (the existence of such a time \( T \) is provided in [14]). Using the Itô–Wentzell formula in Stratonovich form [50] we obtain
\[
dv_t = \left( du_t + \sum_k \xi^{(k)} \cdot \nabla u_t \circ dW^{(k)}_t \right) \bigg|_{x+\sum_k \xi^{(k)} W^{(k)}_t} = du_t \bigg|_{x+\sum_k \xi^{(k)} W^{(k)}_t} + \sum_k \xi^{(k)} \cdot \nabla v_t \circ dW^{(k)}_t. \tag{2.26}
\]

Now, our assumption of constant \( \{\xi^{(k)}\}_{k \in \mathbb{N}} \) implies \( \mathbb{P}(\mathcal{L}^T_{\xi^{(k)}} u_t) = \xi^{(k)} \cdot \nabla u_t \). Thus, using equation (1.5) and (2.26), we obtain the equation, \( dv_t + \mathbb{P}(v_t \cdot \nabla v_t) dt = 0 \).

The classical time derivative \( \partial_t v \) exists since \( \mathbb{P}(v_t \cdot \nabla v_t) \) is continuous-in-time for
each $x$ and, hence, equation (2.25) follows. Note that the change of variables above from $u_t$ to $v_t$ is not a Galilean transformation.

Since the transformation (2.24) is reversible, for sufficiently short times (while solutions exist), the unique stochastic solution $u_t$ of the SPDE (1.5) can be recovered from the unique solution $v_t$ of deterministic Euler (2.25) by evaluating at a random spatial point

$$u_t(x) = v_t \left( x - \sum_k \xi^{(k)} W^{(k)}_t \right).$$

(2.27)

Thus, as discussed in Chapter 5 of [22], no regularizing effects can possibly come by adding this simple multiplicative noise to the Euler equations. If there is any non-trivial regularization-by-noise within the class of Euler–Poincaré models that we consider, it must arise due to spatial variation (and possibly solution dependence) of the noise correlates.

Finally, we mention a related class of models in which the stochasticity is understood to arise from location uncertainty [55, 60–62]. These models also conserve energy pathwise but are distinct from all of those considered here. In particular, they involve an additional division of the fields (velocity and pressure) into slow and fast fluctuating components and are obtained via a version of the Reynold’s transport theorem.

**Stochastic Navier–Stokes–Poincaré equations** In this note, we also obtain a class of natural stochastic generalizations of Navier–Stokes. Similar to the Stochastic Euler–Poincaré equation, we ‘randomize’ the Navier–Stokes equations by insisting that they possess a certain analog of the Kelvin theorem – called the Constantin–Iyer–Kelvin theorem – which we now review. In their paper [9], Constantin and Iyer proved that smooth solutions $u_t$ of the Navier–Stokes equations

$$\partial_t u_t + (u_t \cdot \nabla)u_t = -\nabla p_t + \nu \Delta u_t,$$  \hspace{1cm} (2.28)

$$\nabla \cdot u_t = 0,$$  \hspace{1cm} (2.29)

$$u_t|_{t=0} = u_0,$$  \hspace{1cm} (2.30)

are characterized by the following statistical Kelvin theorem; for all loops $\Gamma \subset \Omega$

$$\int_{\Gamma} u_t \cdot d\ell = \mathbb{E} \left[ \int_{A_t(\Gamma)} u_0 \cdot d\ell \right],$$

(2.31)

where $A_t := X_t^{-1}$ is the back-to-labels map for the stochastic flow defined by the forward Itô equation\(^1\)

$$dX_t(x) = u_t(X_t(x))dt + \sqrt{2\nu} \, dB_t, \quad X_0(x) = x.$$  \hspace{1cm} (2.32)

Here, $B_t$ is a $d$-dimensional standard Brownian motion. The Constantin–Iyer–Kelvin theorem has the beautifully simple implication that smooth Navier–Stokes

\(^1\)Rather than introduce the back-to-labels map, the Constantin–Iyer–Kelvin theorem can also be naturally stated in terms of time-reversed Brownian motion and backwards Itô SDEs [16]. For detailed discussions of backward stochastic flows, see [29, 51].
solutions are uniquely characterized as those velocity fields which have the property that circulations are \textit{backwards martingales} of the stochastic flow (2.32). We note also that F. Rezakhanlou has extended the results of Constantin and Iyer to more irregular drifts \cite{63}, and has advanced a ‘stochastic Hamilton equations’ perspective \cite{64} which is similar in spirit to our work. As we will describe in detail below, theorem 2.12 of the present work identifies a privileged class of stochastic fluid equations characterized by the Lagrangian transport properties discussed in \cite{63,64} along the (doubly) noisy trajectories (2.33). The resulting models can be thought to generalizations of Navier–Stokes to the stochastic setting via the Constantin–Iyer–Kelvin theorem. A few more remarks are now in order.

Unlike the pathwise Kelvin theorem (2.6) which holds for solutions of the Stochastic Euler–Poincaré equations, (2.31) is completely deterministic; since, the fluid velocity \( u_t \) is a solution of equations (2.28)–(2.30). The noise appearing in the flow (2.32) is, in a sense, artificial. It plays a similar role as the noise used in Feynman–Kac representations for linear parabolic equations. Namely, it is a mathematical tool to represent the Laplacian appearing in (2.28). However, unlike the Feynman–Kac representations for linear equations, the stochastic Kelvin theorem (2.31), (2.32) constitutes a nonlinear fixed-point condition since the drift velocity in the trajectories (2.32) is also the solution for which the circulation is computed (2.31). In fact, a stochastic Weber formulation (equivalent to Kelvin theorem for smooth solutions) can be used to prove local existence of solutions of the incompressible Navier–Stokes [44]. See also remark 2.16, below.

We briefly recall some results connected to the formulation (2.31), (2.32). First, a different perspective on the Constantin–Iyer–Kelvin theorem was explored by Eyink in \cite{20}, where it is shown that (2.31) arises as a consequence of Noether’s theorem via the particle relabeling symmetry of a certain stochastic action principle for the deterministic incompressible Navier–Stokes equations. See also \cite{16} for a reformulation of Navier–Stokes as a system of stochastic Hamilton’s equations, which yield a particularly simple derivation of the statistical Kelvin theorem. This formulation has since been extended to domains with solid boundary \cite{10} and to a Riemannian manifold when the de Rham–Hodge Laplacian is the viscous dissipation operator \cite{66}. Finally, Eyink \cite{19} extended the work of Constantin and Iyer to nonideal hydromagnetic models. There, a stochastic analogue of the classical Alfvén theorem was proved to be equivalent to smooth solutions of the deterministic, nonideal, incompressible magnetohydrodynamic equations.

In what follows, we derive a class of SPDEs, smooth solutions of which possess (and are characterized by) a pathwise Constantin–Iyer–Kelvin theorem. We term these the \textit{stochastic Navier–Stokes–Poincaré} equations. Just as for (1.5)–(1.6), these equations are driven by Brownian motions \( \{W_{t}^{(k)}\}_{k \in \mathbb{N}} \) defined on the probability space \( (\Xi, \mathcal{F}, \mathbb{P}) \). Relative to equations (1.5)–(1.6), the stochastic Navier-Stokes–Poincaré equations contain additional terms which can be regarded as arising due to the presence of an artificial Brownian noise on the trajectories, just as in the Constantin–Iyer formalism. This collection of 1-dimensional Brownian motions \( \{B_{t}^{(k)}\}_{k \in \mathbb{N}} \) is independent of the noise \( \{W_{t}^{(k)}\}_{k \in \mathbb{N}} \). We may now state our result.
THEOREM 2.12 (Characterization of Stochastic Navier-Stokes–Poincaré Fluids). Let $X_t$ be the flow defined by
\[
dX_t(x) = u_t(X_t(x))dt + \sum_k \xi^{(k)}(X_t(x)) \circ dW_t^{(k)} + \sqrt{2\nu} \sum_k \eta^{(k)}(X_t(x)) \circ dB_t^{(k)}, \quad X_0(x) = x,
\]
for fixed smooth solenoidal vector fields $u_t: [0, T] \times \Omega \to \mathbb{R}^d$ and $\{\xi^{(k)}\}_{k \in \mathbb{N}}, \{\eta^{(k)}\}_{k \in \mathbb{N}} : \Omega \to \mathbb{R}^d$.

Then, $u_t$ is a smooth solution of equations (2.1) on $[0, T] \times \Omega$ with
\[
\begin{align*}
 f_t &= \mathcal{L}_{\xi^{(k)}}^T u_t - \frac{1}{2} \mathcal{L}_{\xi^{(k)}}^T (\mathcal{L}_{\xi^{(k)}}^T u_t) - \nu \sum_k \mathcal{L}_{\eta^{(k)}}^T (\mathcal{L}_{\eta^{(k)}}^T u_t), \\
 \sigma_t^{(k)} &= \mathcal{L}_{\xi^{(k)}}^T u_t,
\end{align*}
\]
if and only if, for every rectifiable loop $\Gamma$, $u_t$ has the property that for $t \in [0, T]$, conditioned on realizations of $\{W^{(k)}\}_{k \in \mathbb{N}}$, circulations are backwards martingales
\[
\mathbb{E} \left[ \oint_{\Gamma} u_t \cdot d\ell \right] = \mathbb{E} \left[ \mathbb{E} \left[ u_0 \cdot d\ell \bigg| F_t^{W^{(k)}} \right] \bigg| \mathcal{F}_t^{W^{(k)}} \right], \quad \mathbb{P} \text{ a.s.}
\]
where $A_t := X_t^{-1}$ is the back-to-labels map and $\mathcal{F}_t^{W^{(k)}}$ is the sigma-algebra generated by the increments $W_s^{(k)} - W_{s'}^{(k)}$, $0 \leq s' < s \leq t$, $k \in \mathbb{N}$.

The idea above is that, upon conditioning on the knowledge of the processes $\{W^{(k)}\}_{k \in \mathbb{N}}$ during $[0, t]$, we obtain a Constantin–Iyer-type circulation theorem (2.36) by averaging over the ‘unresolved’ Brownian motions $\{B^{(k)}\}_{k \in \mathbb{N}}$. The proof of theorem 2.12 follows a different approach than that of theorem 2.1. Instead of computing the rate of change of circulation and using the Itô–Wentzell formula, we follow the original approach of [9] to prove the equivalence of (2.1) with (2.34) and (2.35) with a fixed-point characterization in terms of a stochastic Weber formula. This, in turn, is equivalent to the Kelvin theorem (2.36).

REMARK 2.13 (Stochastic Fluids with Standard Viscous Friction). If the $B^{(k)}$-noise amplitudes are constant and act only in the $d$ Euclidean directions $\{e_i\}_{i=1}^d$; that is, if
\[
\{\eta^{(k)}\}_{k \in \mathbb{N}} = \{e_1, e_2, e_3, \ldots, e_d, 0, 0, 0, \ldots\},
\]
then $\nu \sum_k \mathcal{L}_{\eta^{(k)}}^T (\mathcal{L}_{\eta^{(k)}}^T u_t) = \nu \Delta u_t$ and the usual viscous Laplacian appearing in (2.28) is recovered. Thus (2.1) with (2.34) and (2.35) and $\{\eta^{(k)}\}_{k \in \mathbb{N}}$ given by (2.37) form a family of stochastic generalizations of the deterministic Navier–Stokes which satisfy the Constantin–Iyer–Kelvin relation (2.36).

REMARK 2.14 (Energetic Properties of Circulation–Theorem Preserving Stochastic Fluids). We now consider the energetics of the stochastic circulation–theorem–preserving models discussed here. Using (2.1) with (2.34) and (2.35) (the case with
Recall from (A.5) in Appendix A that the Lie derivative of a vector field $w$ is defined by

$$-\mathcal{L}_{\xi} w = [\xi, w] := \xi \cdot \nabla w - w \cdot \nabla \xi,$$

and its adjoint operator satisfies the identity $(\mathcal{L}_{\xi}^T v, w)_{L^2(\Omega)} = -\langle v, \mathcal{L}_{\xi} w \rangle_{L^2(\Omega)}$, see equation (A.8). Upon integrating by parts in (2.38) using the adjoint relation and recalling that $u_t$ is divergence-free, we find

$$d\|u_t\|_{L^2(\Omega)}^2 = -\frac{1}{2} \sum_k (\mathcal{L}_{\xi(k)} u_t, \mathcal{L}_{\xi(k)}^T u_t)_{L^2(\Omega)} dt - \nu \sum_k (\mathcal{L}_{\eta(k)} u_t, \mathcal{L}_{\eta(k)}^T u_t)_{L^2(\Omega)} dt$$

$$+ \frac{1}{2} \sum_k (P \mathcal{L}_{\xi(k)}^T u_t, \mathcal{L}_{\xi(k)}^T u_t)_{L^2(\Omega)} dt + \sum_k (u_t, \nabla \xi^{(k)} \cdot u_t)_{L^2(\Omega)} dW^k_t.$$

(2.39)

Now, if $\xi$ and $v$ are divergence-free, then so is $\mathcal{L}_{\xi} v$. Consequently, we find that

$$(\mathcal{L}_{\xi(k)} u_t, \mathcal{L}_{\xi(k)}^T u_t)_{L^2(\Omega)} = (P \mathcal{L}_{\xi(k)}^T u_t, \mathcal{L}_{\xi(k)}^T u_t)_{L^2(\Omega)}$$

$$= (P(\mathcal{L}_{\xi(k)} u_t - \mathcal{L}_{\xi(k)}^T u_t), \mathcal{L}_{\xi(k)}^T u_t)_{L^2(\Omega)}$$

$$= -(P(u_t \cdot \nabla \xi^{(k)} + \nabla \xi^{(k)} \cdot u_t), \mathcal{L}_{\xi(k)}^T u_t)_{L^2(\Omega)}.$$

Thus

$$d\|u_t\|_{L^2(\Omega)}^2 = \frac{1}{2} \sum_k (P(u_t \cdot \nabla \xi^{(k)} + \nabla \xi^{(k)} \cdot u_t), \mathcal{L}_{\xi(k)}^T u_t)_{L^2(\Omega)} dt$$

$$- \nu \sum_k (\mathcal{L}_{\eta(k)} u_t, \mathcal{L}_{\eta(k)}^T u_t)_{L^2(\Omega)} dt + \sum_k (u_t, \nabla \xi^{(k)} \cdot u_t)_{L^2(\Omega)} dW^k_t.$$

(2.40)

Unlike equations (2.1) with (2.21) and (2.22) discussed in remark 2.10, the above computation shows that circulation-theorem preserving models do not, in general, satisfy a simple energy equality even when $\nu \equiv 0$ unless the $\xi^{(k)}$ are spatially constant. Firstly, the energy in (2.38) is a fluctuating quantity. Moreover, even the average energy is neither increasing, nor decreasing, a priori. Energy can be introduced or removed from the system by the action of spatial gradients of the noise correlates $\{\xi^{(k)}\}$ on the solution. However it is clear from (2.39) that if, for example,
the $\eta^{(k)}$ are constant in space and span $\mathbb{R}^d$ (e.g. as in remark 2.13) and if $\nu$ is taken to be sufficiently large, relative to the magnitude of $\xi^{(k)}$ and its spatial gradient, then the system is dissipative on the average. That is, smooth solutions satisfy the inequality
\[
\mathbb{E} \left\| \frac{1}{2} \| u_t \|_{L^2(\Omega)}^2 \right\| \leq \frac{1}{2} \| u_0 \|_{L^2(\Omega)}^2,
\]
where the expectation $\mathbb{E}$ denotes averaging over the Brownian motions $\{W_t^{(k)}\}_{k \in \mathbb{N}}$. Thus, among the class of models (2.1) satisfying (2.34) and (2.35) (i.e., among the choices for $\xi^{(k)}$), there are equations which have solutions possessing the Constantin–Iyer Kelvin theorem $\mathcal{P}$ almost surely and are, on the average, dissipative.

**Remark 2.15 (Energetics of Dissipating Stochastic Fluids).** We describe one last class of models; those which dissipate energy pathwise and thus generalize (2.1) with (2.21) and (2.22) to the non-ideal setting. Fixing solenoidal vector fields $\{\xi^{(k)}\}_{k \in \mathbb{N}}$ and $\{\eta^{(k)}\}_{k \in \mathbb{N}}$ and using the notation introduced for (2.20), they read
\[
du_t + B(u_t)dt + \sum_k \xi^{(k)} \circ dW_t^{(k)}, u_t) = \nu \sum_k \mathbb{P}(\eta^{(k)} \cdot \nabla \eta^{(k)} \cdot \nabla u_t)).
\]
(2.42)
The form of the ‘viscous term’ is chosen as the piece of the double-(adjoint) Lie operator $\mathcal{L}^{(k)}_t(\mathcal{L}^{(k)}_t u_t)$ appearing in the stochastic Navier–Stokes–Poincaré equations which ensures that this term cannot increase the energy. There are, of course, other choices for the dissipation operator. In Itô form, equation (2.42) is (2.1) with
\[
f_t := u_t \cdot \nabla u_t - \sum_k \xi^{(k)} \cdot \nabla \mathbb{P}(\eta^{(k)} \cdot \nabla u_t) - \nu \sum_k \eta^{(k)} \cdot \nabla \mathbb{P}(\eta^{(k)} \cdot \nabla u_t),
\]
(2.43)
\[
\sigma^{(k)}_t := \xi^{(k)} \cdot \nabla u_t.
\]
(2.44)
Due to the properties of $B(w, v)$ discussed in remark 2.10, solutions of (2.1) with (2.43) and (2.44) satisfy an pathwise energy balance
\[
\frac{1}{2} \| u_t \|_{L^2(\Omega)}^2 = \frac{1}{2} \| u_0 \|_{L^2(\Omega)}^2 - \nu \sum_k \int_0^t \| \mathbb{P}(\eta^{(k)} \cdot \nabla u_s) \|_{L^2(\Omega)}^2 ds, \quad \mathcal{P} \text{ a.s.}
\]
(2.45)
Unsurprisingly, such fluids do not possess a Constantin–Iyer–Kelvin theorem, in general, unless the noise vector fields $\{\xi^{(k)}\}_{k \in \mathbb{N}}$ and $\{\eta^{(k)}\}_{k \in \mathbb{N}}$ are spatially constant.

**Remark 2.16 (Pathwise Stochastic Weber Formula).** In the proof of theorem 2.12, we show that solutions of the stochastic Navier–Stokes–Poincaré equations (1.5)–(1.6) satisfy a pathwise stochastic Weber formula:
\[
u^s_t(x) = \mathbb{E}[\mathbb{P}[\nabla A_t(x)^T u_0(A_t(x))]^2 \bigg| \mathcal{F}^{W^{(k)}}_t], \quad \mathcal{P} \text{ a.s.}
\]
(2.46)
in which the expectation averages over the standard Brownian motions $\{B^{(k)}\}_{k \in \mathbb{N}}$. By Stokes theorem applied to (2.36), we find that the vorticity-flux through
comoving areas is statistically frozen

\[ \int \int \omega_t \cdot dS = \mathbb{E} \left[ \int \int_{A_t(S)} \omega_0 \cdot dS \bigg| \mathcal{F}_t^W \right], \quad \text{P a.s.} \quad (2.47) \]

3. Proofs

Proof of proposition 2.5. The proof follows from a direct computation. First, we convert (2.9) to an equivalent Itô SDE governing the paths

\[ dX_t(x) = \left( u_t + \frac{1}{2} \sum_k \xi^{(k)} \cdot \nabla \xi^{(k)} \right) \bigg|_{X_t(x)} dt + \sum_k \xi^{(k)}(X_t(x)) dW_t^{(k)}, \quad X_0(x) = x. \quad (3.1) \]

The new term appearing in (3.1) is called the ‘noise-induced drift’. Now, for any rectifiable loop \( \Gamma \), let \( \Gamma(s) : [0, 1] \to \Gamma \) be a parametrization. Then the circulation around the loop \( \Gamma \) can be represented as

\[ \oint_{X_t(\Gamma)} u_t \cdot d\ell = \int_0^1 \frac{d}{ds} X_t(\Gamma(s)) \cdot u_t(X_t(\Gamma(s))) ds \]

\[ = \int_0^1 \Gamma'(s) \cdot \nabla X_t(\Gamma(s)) \cdot u_t(X_t(\Gamma(s))) ds. \quad (3.2) \]

Upon differentiating the circulation in this representation and applying the Itô product rule, we have

\[ d \oint_{X_t(\Gamma)} u_t \cdot d\ell = \int_0^1 \Gamma'(s) \cdot (\nabla X_t \cdot du_t(X_t)) \bigg|_{x=\Gamma(s)} ds \]

\[ + d[\nabla X_t \cdot u_t(X_t)] + d[\nabla X_t; u_t(X_t)] \bigg|_{x=\Gamma(s)} ds. \quad (3.3) \]

The flow \( u_t \) is random, driven by the same noise as on the particle trajectories. Therefore, to compute the stochastic differential \( d(u_t(X_t(x))) \), we apply the Itô–Wentzell formula. For details, see, e.g., theorem 1.1. of [50] or theorem 3.3.1 of [51]. This calculation introduces the Wentzell correction, as

\[ d(u_t(X_t(x))) = (du_t + dX_t \cdot \nabla u_t) \bigg|_{X_t(x)} \]

\[ + \frac{1}{2} \nabla \otimes \nabla u_t : d[X_t, X_t]_{t} + d[\nabla u_t; X_t]_{t} \bigg|_{X_t(x)} \]

\[ = \left( du_t + \left( u_t \cdot \nabla u_t + \frac{1}{2} \sum_k (\xi^{(k)} \cdot \nabla) \xi^{(k)} \cdot \nabla u_t \right) \right) \bigg|_{X_t(x)} \]

\[ + \frac{1}{2} \sum_k \xi^{(k)} \otimes \xi^{(k)} : \nabla \otimes \nabla u_t \right)\bigg|_{X_t(x)} \]

\[ + \sum_k (\xi^{(k)} \cdot \nabla) u_t \bigg|_{X_t(x)} dW_t^{(k)} + d[\nabla u_t; X_t]_{t} \bigg|_{X_t(x)}. \]
To compute the Wentzell correction \( d[\nabla u_t; X_t] \big|_{X_t(x)} \) explicitly, we take the gradient of equation (2.1)
\[
d\nabla u_t = -\nabla P \cdot f_t dt - \sum_k \nabla \sigma_t^{(k)} dW_t^{(k)}.
\] (3.4)

The martingale part of \( d\nabla u_t \) is \( -\sum_k \nabla \sigma_t^{(k)} dW_t^{(k)} \). Consequently, the Wentzell correction is given by
\[
d\big|_{\nabla u_t; X_t} \big|_{X_t(x)} = d\big[ \partial_i u_t, X^i_t \big] = -\sum_k (\xi^{(k)} \cdot \nabla) \sigma_t^{(k)} \big|_{X_t(x)} dt. \] (3.5)

Putting this together, we obtain the full differential
\[
d(u_t(X_t(x))) = (u_t \cdot \nabla u_t - \nabla f_t) \big|_{X_t(x)} dt + \sum_k ((\xi^{(k)} \cdot \nabla) u_t - \nabla \sigma_t) \big|_{X_t(x)} dW_t^{(k)}
\]
\[+ \sum_k \left( \frac{1}{2} (\xi^{(k)} \cdot \nabla) \xi^{(k)} \cdot \nabla u_t + \frac{1}{2} \xi^{(k)} \otimes \xi^{(k)} : \nabla \otimes \nabla u_t - \xi^{(k)} \cdot \nabla \sigma_t \right) \big|_{X_t(x)} dt. \] (3.6)

Next, the gradient of the stochastic flow is easily found to satisfy
\[
d\nabla X_t(x) = \nabla X_t(x) \cdot \left( \nabla u_t(X_t(x)) + \frac{1}{2} \sum_k \nabla (\xi^{(k)} \cdot \nabla \xi^{(k)}) \right) dt \]
\[+ \sum_k \nabla X_t(x) \cdot \nabla \xi^{(k)}(X_t(x)) dW_t^{(k)}, \]
\[\nabla X_0(x) = I. \] (3.8)

In view of (3.6) and (3.7), the quadratic cross-variation between the Lagrangian velocity and deformation matrix is
\[
d[\nabla X_t; u_t(X_t)] \big|_t = \nabla X_t \cdot \sum_k \nabla \xi^{(k)} \cdot ((\xi^{(k)} \cdot \nabla) u_t - \nabla \sigma_t) \big|_{X_t(x)} dt. \] (3.9)

Finally, the remaining term in (3.3) can be expressed using (3.7) as follows
\[
d\nabla X_t \cdot u_t(X_t) = \nabla X_t \cdot \left( \left( \nabla \left( \frac{1}{2} |u_t|^2 \right) + \frac{1}{2} \sum_k \nabla (\xi^{(k)} \cdot \nabla \xi^{(k)}) \cdot u_t \right) dt \right.
\]
\[+ \sum_k (\nabla \xi^{(k)} \cdot u_t) dW_t^{(k)} \bigg|_{X_t(x)} \bigg]. \] (3.10)
Upon putting the various elements of this computation together, we have
\[
\int_{X_t(\Gamma)} u_t \cdot d\ell = \int_{X_t(\Gamma)} \left( L^T_{u_t} u_t - f_t \right) dt \cdot d\ell \\
+ \sum_k \int_{X_t(\Gamma)} \left( \frac{1}{2} (\xi(k) \cdot \nabla) \xi(k) \cdot \nabla u_t + \frac{1}{2} \xi(k) \otimes \xi(k) \cdot \nabla \otimes \nabla u_t \\
+ \frac{1}{2} (\xi(k) \cdot \nabla \xi(k)) : u_t + \nabla (\xi(k) \cdot \nabla) u_t - L^T_{\xi(k)} \sigma_t^{(k)} \right) dt \cdot d\ell \\
+ \sum_k \int_{X_t(\Gamma)} \left( L^T_{\xi(k)} u_t - \sigma_t \right) dW_t^{(k)} \cdot d\ell.
\]

Recall from the computation (2.8) that \( L^T_\xi P v = L^T_\xi v + \nabla (\xi \cdot \nabla q) \), for any vector field \( v \) and some scalar function \( q \). Since gradients vanish upon integration over closed loops (and, consequently, the action of the Leray projector is trivial on loop integrals), we have that
\[
d \int_{X_t(\Gamma)} u_t \cdot d\ell = \int_{X_t(\Gamma)} \left( L^T_{u_t} u_t - f_t \right) dt \cdot d\ell \\
+ \sum_k \int_{X_t(\Gamma)} \left( \frac{1}{2} (\xi(k) \cdot \nabla) \xi(k) \cdot \nabla u_t + \frac{1}{2} \xi(k) \otimes \xi(k) \cdot \nabla \otimes \nabla u_t \\
+ \frac{1}{2} (\xi(k) \cdot \nabla \xi(k)) : u_t + \nabla (\xi(k) \cdot \nabla) u_t - L^T_{\xi(k)} \sigma_t^{(k)} \right) dt \cdot d\ell \\
+ \sum_k \int_{X_t(\Gamma)} \left( L^T_{\xi(k)} u_t - \sigma_t \right) dW_t^{(k)} \cdot d\ell.
\]

Now note that the double (adjoint) Lie derivative (1.6) can be expanded as follows:
\[
L^T_\xi (L^T_\xi v) = L^T_\xi (\xi^j \partial_j v_i + \partial_i \xi^j v_j) \\
= \xi^k \partial_k (\xi^j \partial_j v_i + \partial_i \xi^j v_j) + \partial_i \xi^k (\xi^j \partial_j v_k + \partial_k \xi^j v_j) \\
= (\xi \cdot \nabla) \xi \cdot \nabla v_i + (\xi \otimes \xi) : (\nabla \otimes \nabla) v \\
+ \partial_i \xi^j (\xi \cdot \nabla) v_j + ((\xi \cdot \nabla) \partial_i \xi^j) v_j + (\partial_i \xi^j)(\xi \cdot \nabla) v_j + (\partial_i \xi^k) \partial_k \xi^j v_j \\
= (\xi \cdot \nabla) \xi \cdot \nabla v + (\xi \otimes \xi) : (\nabla \otimes \nabla) v + 2 \nabla \xi \cdot (\xi \cdot \nabla) v + \nabla ((\xi \cdot \nabla) \xi) \cdot v.
\]

Upon substituting this simplification into (3.12), we finally obtain equation (2.10).
We remark that, geometrically, the right-hand side of equation (3.13) is the $L^2$ dual of the double Lie bracket $[\xi, [\xi, u]]$ of the vector field $\xi$ acting on $u$; that is,
$$\langle \mathcal{L}_\xi^T (\mathcal{L}_\xi^T v), u \rangle = \langle v, [\xi, [\xi, u]] \rangle.$$ 
Moreover, in three-dimensional Euclidean space, by using the second form given in equation (1.6) one can obtain the following alternative expression for the double (adjoint) Lie derivative involving cross-products and the curl operator:
$$\mathcal{L}_\xi^T (\mathcal{L}_\xi^T v) = \mathcal{L}_\xi^T (-\xi \times \text{curl} + \nabla(\xi \cdot v))$$
$$= \xi \times \text{curl}(\xi \times \text{curl} v) - \nabla(\xi \cdot (\xi \times \text{curl} v)) + \mathcal{L}_\xi^T (\nabla(\xi \cdot v))$$
$$= \xi \times \text{curl}(\xi \times \text{curl} v) + \mathcal{L}_\xi^T (\nabla(\xi \cdot v))$$
$$= \xi \times \text{curl}(\xi \times \text{curl} v) + \nabla(\xi \cdot \nabla(\xi \cdot v)), \quad (3.14)$$
where we have used the identity $\xi \cdot (\xi \times \text{curl} v) = \text{curl} v \cdot (\xi \times \xi) = 0$ and the fact that $\mathcal{L}_\xi^T \nabla q = \nabla(\xi \cdot \nabla q)$ which was verified in equation (2.8). Note that the final term in (3.14) is a total gradient and therefore vanishes upon integration over any closed, rectifiable loop $\Gamma$.

**Proof of Theorem 2.1.** We proceed in the same spirit as in the proof of the equivalence of the usual Kelvin theorem to smooth solutions of deterministic Euler given in the Introduction.

**Direction 1: Stochastic Euler–Poincaré solutions have a pathwise Kelvin theorem.** In view of proposition 2.5, one direction is simple: by using equation (2.1) with $f_t$ and $\{\sigma_{t}^{(k)}\}_{k \in \mathbb{N}}$ defined by (2.4) and (2.5) in theorem 2.1, and applying proposition 2.5 to the unique smooth solution $u_t$ for given initial conditions $u_0$ (which always exists provided, at least, that $T$ is taken sufficiently small \[14, 25\], see remark 2.4), one has that realization-by-realization of the Brownian processes $\{W_t^{(k)}\}_{k \in \mathbb{N}}$ circulations are materially conserved (2.6).

**Direction 2: Pathwise Kelvin theorem for all loops implies $u_t$ is a stochastic Euler–Poincaré solution.** For the other direction, assume that the circulation is conserved along all material loops $\Gamma$. Since $u_t$ and $\{\xi^{(k)}\}_{k \in \mathbb{N}}$ are assumed smooth, the map $x \mapsto X_t(x)$ is a $\mathcal{F}_t$-adapted diffeomorphism \[51, 52\]. Its spatial inverse $A_t = X_t^{-1}$ is $\mathcal{F}_t$-adapted, pointwise in $x$. See remark 2.3 for a precise, sufficient regularity assumption. First we establish the form of the noise in the SPDE.

**Form of noise:** From (2.10), the quadratic variation of the circulation (denoted for a process $\zeta_t$ by $[\zeta_t]_t$) is
$$\left[ \int_{X_t(\Gamma)} u_t \cdot d\ell \right]_{t=T'}^{t=T} = \sum_{k} \int_{0}^{T'} \int_{0}^{1} \left| \Gamma'(s) \cdot \nabla X_t(\Gamma(s)) \cdot (\mathcal{L}^T_{\xi^{(k)}} u_t - \sigma_t^{(k)}) \right|^2 \, dt \, ds, \quad (3.15)$$
for any $T' \in [0, T]$. On the other hand, if the pathwise Kelvin theorem holds, then the left-hand side must vanish. By assumption, the function $f(t, s) :=$
Theorem preserving stochastic fluids

\[ \Gamma'(s) \cdot \nabla X_t(\Gamma(s)) \cdot (L^T_{\xi(k)}u_t - \sigma_t^{(k)})|_{X_t(\Gamma(s))} \text{ is continuous on } [0, T] \times [0, 1]. \] Thus, we conclude that for all \( (t, s) \in [0, T] \times [0, 1], \)

\[ \Gamma'(s) \cdot \nabla X_t(\Gamma(s)) \cdot (L^T_{\xi(k)}u_t - \sigma_t^{(k)})|_{X_t(\Gamma(s))}dt = 0, \quad \forall \ k \in \mathbb{N}. \quad (3.16) \]

We now show that the matrix \( \nabla X_t \) in (3.16) is non-singular almost surely for all \( x \in \Omega \). For this, we apply

**Lemma 3.1.** Fix smooth vector fields \( b_t : [0, T] \times \Omega \to \mathbb{R}^d \) and \( \{\xi^{(k)}\}_{k \in \mathbb{N}} : \Omega \to \mathbb{R}^d \). Let \( x \mapsto X_{s,t}(x) \) be the regular stochastic flow of diffeomorphisms [51] associated to the Itô SDE

\[ dX_t(x) = b_t(X_t(x))dt + \sum_k \xi^{(k)}(X_t(x))dW_t^{(k)}, \quad X_0(x) = x. \quad (3.17) \]

Then the following formula for the Jacobian holds

\[ \det(\nabla X_t(x)) = \exp \left( \int_0^t \left( \nabla \cdot b_t - \frac{1}{2} \sum_k (\nabla \xi^{(k)} )^T : \nabla \xi^{(k)} \right) |_{X_s(x)} ds \right. \]

\[ \left. + \sum_k \int_0^t \nabla \cdot \xi^{(k)}|_{X_s(x)} dW_s^{(k)} \right). \quad (3.18) \]

**Proof.** Recall the classic formula \( \ln(\det A) = \text{tr}(\ln A) \), for any invertible matrix \( A \). The first- and second-order Gateaux derivative of \( \ln(\det A) \) in direction \( \phi \) and in \((\phi, \psi)\) resp. may then be computed to be

\[ D \ln(\det A)[\phi] = \text{tr}[\phi A^{-1}], \quad D^2 \ln(\det A)(A)[\phi, \psi] = -\text{tr}[\phi A^{-1} \psi A^{-1}]. \quad (3.19) \]

The proof will follow as a direct computation. First, by taking the gradient in the initial data of (3.17) we have

\[ d\nabla X_t(x) = \nabla X_t(x) \cdot \nabla b_t(X_t(x))dt \]

\[ + \sum_k \nabla X_t(x) \cdot \nabla \xi^{(k)}(X_t(x))dW_t^{(k)}, \quad \nabla X_0(x) = \mathbb{I}. \quad (3.20) \]
Next, applying Itô’s formula and using equation (3.20) and the formulae (3.19), we compute

\[
\begin{align*}
d \ln \det(\nabla X_t(x)) \\
= D \ln \det(\nabla X_t(x)) \nabla X_t(x) \cdot \nabla b_t(X_t(x)) dt \\
+ \sum_k D \ln \det(\nabla X_t(x)) \nabla X_t(x) \cdot \nabla \xi^{(k)}(X_t(x)) dW_t^{(k)} \\
+ \frac{1}{2} \sum_k D^2 \ln \det(\nabla X_t(x)) \nabla X_t(x) \cdot \nabla \xi^{(k)}(X_t(x)) dW_t^{(k)} \\
= \text{tr}(\nabla b_t(X_t(x))) dt + \sum_k \text{tr}(\nabla \xi^{(k)})(X_t(x)) dW_t^{(k)} \\
- \frac{1}{2} \sum_k (\nabla \xi^{(k)})^T : \nabla \xi^{(k)}|_{X_t(x)} dt.
\end{align*}
\]

We integrate in time and evaluate \( \ln \det(\nabla X_0(x)) = 0, \) since \( \det(\nabla X_0(x)) = 1. \) This yields a formula for \( \ln \det(\nabla X_t(x)) \); whereupon formula (3.18) emerges, upon exponentiating the result. \( \square \)

In view of (3.1), we apply lemma 3.1 with \( b_t = u_t + 1/2 \sum_k \xi^{(k)} \cdot \nabla \xi^{(k)} \). Note that

\[
\nabla \cdot b_t = \nabla \cdot u_t + \frac{1}{2} \sum_k \left( (\nabla \xi^{(k)})^T : \nabla \xi^{(k)} + \sum_k \xi^{(k)} \cdot \nabla (\nabla \cdot \xi^{(k)}) \right).
\]

Thus, for divergence-free vector fields \( u_t \) and \( \{\xi^{(k)}\}_{k \in \mathbb{N}} \), we find from (3.18) that the Stratonovich stochastic flow (2.3) is volume preserving, \( \det(\nabla X_t(x)) = 1. \) Thus, the kernel of \( \nabla X_t \) is trivial \( \mathbb{P} \) almost surely pointwise in \( (t, x) \in [0, T] \times \Omega \). Now, for any point \( x \in \Omega \), choose a collection of loops \( \{\Gamma_i(s)\}_{i=1, \ldots, d} \) such that at \( x = \Gamma_i(s_i) \) for some \( s_i \in [0, 1] \) and with linearly independent tangents \( \{\Gamma'_i(s_i)\}_{i=1, \ldots, d} \). Since (3.16) holds for all such loops and the matrix \( \nabla X_t(x) \) is non-singular, it follows that 

\[
\mathcal{L}_{\xi}^T u_t - \sigma_t^{(k)} = 0 \quad \text{at } X_t(x) \quad \text{for all } t \in [0, T], \quad \mathbb{P} \text{ almost surely.}
\]

The above argument can be applied to all \( x \in \Omega \) by choosing the appropriate collection of loops \( \{\Gamma_i\} \) and we conclude,

\[
\sigma_t^{(k)}|_{X_t(x)} = \mathcal{L}_{\xi}^T u_t|_{X_t(x)}, \quad \forall k \in \mathbb{N}, \quad (t, x) \in [0, T] \times \Omega, \quad \mathbb{P} \text{ a.s.} \tag{3.21}
\]

Finally, fix any \( y \in \Omega \). Then, for any \( t \in [0, T] \) and \( \mathbb{P} \) a.e. \( \varpi \) (where \( \varpi \) denotes sample space dependence), letting \( x = A_t^\varpi(y) \) allows us to conclude that \( \sigma_t^{(k)} = \mathcal{L}_{\xi}^T u_t \) for all \( (t, y) \in [0, T] \times \Omega, \) \( \mathbb{P} \) almost surely.

**Form of drift.** Upon using the fact that \( \sigma_t^{(k)} = \mathcal{L}_{\xi}^T u_t \), pointwise in spacetime \( \mathbb{P} \) a.s., Prop. 2.5, implies that

\[
\int_0^T \int_{X_t(\Gamma)} (\mathcal{L}_{u_t}^T u_t - \sum_k \frac{1}{2} \mathcal{L}_{\xi}^T (\mathcal{L}_{\xi}^T u_t) - f_t) \cdot d\ell \, dt = 0,
\]
for all rectifiable loops $\Gamma$ and all $T' \in [0, T]$. Since it is continuous, the integrand in the time integral above must vanish identically for all $t \in [0, T]$. Now, let $\Gamma'$ be any rectifiable loop. Then, for any fixed $t \in [0, T]$ and $P$ a.e. $\varpi$, let $\Gamma = A^P_t(\Gamma')$. Thus, we deduce that for any loop $\Gamma'$ the following holds

$$\oint_{\Gamma'} \left( L^T_{u_t} u_t - \sum_k \frac{1}{2} L^T_{\xi(k)} (L^T_{\xi(k)} u_t) - f_t \right) \cdot d\ell = 0.$$ 

Finally, we can conclude that there exists a scalar process $q_t$ (not necessarily of bounded variation) such that

$$f_t = \left( L^T_{u_t} u_t - \sum_k \frac{1}{2} L^T_{\xi(k)} (L^T_{\xi(k)} u_t) \right) + \nabla q_t.$$ 

By the fact that the Leray–Hodge projector $P$ vanishes on gradients, it follows that equation (2.1) is satisfied with $f_t$ given by the expression (2.4). □

**Proof of theorem 2.12.** Our proof employs a different method than that of our theorem 2.1. In particular, we establish equivalence to a stochastic Weber formula,

$$u_t(x) = \mathbb{E}[\mathbb{P}(\nabla A_t(x))^T u_0(A_t(x)) \bigg| \mathcal{F}^{W^{(k)}}_t], \ P \ a.s. \quad (3.22)$$

where $A_t = X_t^{-1}$ is the back-to-labels map and $X_t$ solves (2.33). Note that, together, equations (3.22) and (2.33) form a fixed point problem. It should be possible to solve this problem (pathwise in $W^{(k)}$) by combining the methods of [25] for the stochastic Euler–Poincaré with those of [44] which establish local existence of deterministic Navier–Stokes from the stochastic Weber formula. We do not pursue this issue here. Instead, we simply assume that smooth solutions of (2.33), (3.22) exist, at least for sufficiently small times $T > 0$.

Note that it is clear that for sufficiently smooth $u_t$, the stochastic Weber formula (3.22) is equivalent to its integrated form on loops – the Constantin–Iyer Kelvin theorem:

$$\oint_{\Gamma} u_t \cdot d\ell = \mathbb{E} \left[ \oint_{\Gamma} (\nabla A_t(x))^T u_0(A_t(x)) \bigg| \mathcal{F}^{W^{(k)}}_t \right] \cdot d\ell = \mathbb{E} \left[ \oint_{A_t(\Gamma)} u_0 \cdot d\ell \bigg| \mathcal{F}^{W^{(k)}}_t \right].$$

Thus, equivalence to the Constantin–Iyer Kelvin theorem for smooth solutions will follow from the same fixed point problem and the stochastic Navier–Stokes–Poincaré equations (2.1) with $f_t$ and $\{\sigma^k_t\}_{k \in \mathbb{N}}$ defined by (2.34) and (2.35). We note that this strategy has also been used in [19] to prove the equivalence of certain non-ideal hydromagnetic models to their stochastic Alfvén theorems.

**Direction 1: Solution of the fixed-point problem (2.33), (3.22) solves equation (2.1).** We first prove that a solution of the fixed point problem (3.22) provides a representation for a solution of equation (2.1) with (2.34) and (2.35).
We begin by using (3.22) to show that for any solenoidal vector field \( v \) for all \( 0 \leq s \leq t \leq T \) we have

\[
\langle u_t, v \rangle_{L^2} = \mathbb{E}[\langle (\nabla A_t)^T u_0(A_t), v \rangle_{L^2}] = \mathbb{E}[\langle u_0(A_t), (\nabla A_t)v \rangle_{L^2}]
\]

\[
= \mathbb{E}[\langle u_0, (\nabla A_t)(X_t)v(X_t) \rangle_{L^2}] = \mathbb{E}[\langle u_0, (A_t)^*v \rangle_{L^2}]
\]

(3.23)

where we have recalled that \((\nabla A_t)(X_t)v(X_t) := (A_t)^*v = (X_t^{-1})^*v\) is the pull-back of \( v \) by the flow \( X_t \). Now, by Kunita’s formula [51], we have for flows \( X_t \) generated by the SDE (2.33) that

\[
(A_t)^*v = v + \sum_k \int_0^t (A_s)^*\mathcal{L}_{\xi(k)} v \ dW_s^{(k)} + \sqrt{2\nu} \sum_k \int_0^t (A_s)^*\mathcal{L}_{\eta(k)} v \ dB_s^{(k)}
\]

\[
+ \int_0^t \left[ (A_s)^*\mathcal{L}_{u_s} v + \frac{1}{2} \sum_k (A_s)^*\mathcal{L}_{\xi(k)}(\mathcal{L}_{\xi(k)} v) \right] \, ds.
\]

(3.24)

In the interest of being self-contained, we prove the identity (3.24) in a slightly different but equivalent form in lemma 3.3 below. Substituting (3.24) into (3.23) and recalling that \( \sqrt{2\nu} \sum_k \int_0^t (A_s)^*\mathcal{L}_{\eta(k)} v \ dB_s^{(k)} \) is a martingale, conditioned on the history of the process \( W_t^{(k)} \), we have

\[
\langle u_t, v \rangle_{L^2} = \langle u_0, v \rangle_{L^2} + \sum_k \mathbb{E} \left[ \int_0^t \langle u_0, (A_s)^*\mathcal{L}_{\xi(k)} v \rangle_{L^2(\Omega)} \ dW_s^{(k)} \right]
\]

\[
+ \int_0^t \mathbb{E} \left[ \langle u_0, (A_s)^*\mathcal{L}_{u_s} v \rangle_{L^2(\Omega)} + \frac{1}{2} \sum_k \langle u_s, (A_s)^*\mathcal{L}_{\xi(k)}(\mathcal{L}_{\xi(k)} v) \rangle_{L^2(\Omega)} \right]
\]

\[
+ \nu \sum_k \langle u_0, (A_s)^*\mathcal{L}_{\eta(k)}(\mathcal{L}_{\eta(k)} v) \rangle_{L^2(\Omega)} \right] \, ds.
\]

(3.25)

Upon using the equivalence (3.23), which holds for any divergence-free vector field (a property which is satisfied by all of \( \mathcal{L}_{u_s} v, \mathcal{L}_{\xi(k)} v, \mathcal{L}_{\xi(k)}(\mathcal{L}_{\xi(k)} v), \) and \( \mathcal{L}_{\eta(k)}(\mathcal{L}_{\eta(k)} v) \) since \( \xi(k) \) and \( \eta(k) \) are assumed solenoidal), we see that

\[
\langle u_t, v \rangle_{L^2} = \langle u_0, v \rangle_{L^2} + \sum_k \int_0^t \langle u_s, \mathcal{L}_{\xi(k)} v \rangle_{L^2(\Omega)} \ dW_s^{(k)}
\]

\[
+ \int_0^t \left[ \langle u_s, \mathcal{L}_{u_s} v \rangle_{L^2(\Omega)} + \frac{1}{2} \sum_k \langle u_s, \mathcal{L}_{\xi(k)}(\mathcal{L}_{\xi(k)} v) \rangle_{L^2(\Omega)} \right]
\]

\[
+ \nu \sum_k \langle u_s, \mathcal{L}_{\eta(k)}(\mathcal{L}_{\eta(k)} v) \rangle_{L^2(\Omega)} \right] \, ds.
\]

(3.26)
The resulting equation corresponds exactly with the definition of the weak form (2.2), thereby establishing that (3.22) is the solution in the sense of definition 3 of [14].

**Direction 2:** Smooth solutions equation (2.1) satisfy the fixed-point problem (2.33), (3.22). Given a smooth solution \( u_t \), we may construct a smooth flow \( X_t \) solving (3.22), as well as its back-to-labels map \( A_t \) which solves

\[
\begin{align*}
\frac{d}{dt} A_t(x) + u_t(x) \cdot \nabla A_t(x) dt + \sum_k \xi^{(k)} \cdot \nabla A_t(x) \circ dW_t^{(k)} \\
+ \sqrt{2\nu} \sum_k \eta^{(k)} \cdot \nabla A_t(x) \circ dB_t^{(k)} &= 0,
\end{align*}
\]

with data \( A_t(x)|_{t=0} = x \). This equation is easily established by applying the Itô formula to \( A_t \circ X_t = \text{id} \). The spatial gradient of the back-to-labels map is then found to solve

\[
\begin{align*}
\frac{d}{dt} \nabla A_t(x) + \mathcal{L}_{u_t} \nabla A_t(x) dt + \sum_k \mathcal{L}_{\xi^{(k)}} \nabla A_t(x) \circ dW_t^{(k)} \\
+ \sqrt{2\nu} \sum_k \mathcal{L}_{\eta^{(k)}} \nabla A_t(x) \circ dB_t^{(k)} &= 0,
\end{align*}
\]

with data \( \nabla A_t(x)|_{t=0} = \mathbb{I} \). Define now \( \bar{u}_t := \bar{u}_t(x) \) from \( u_0, A_t \) and \( \nabla A_t \) by

\[
\bar{u}_t(x) = \mathbb{E} \left[ \mathbb{P}(\nabla A_t(x))^T u_0(A_t(x)) \left| \mathcal{F}_t^{(W^{(k)})} \right] , \quad \mathbb{P} \text{ a.s.} \tag{3.29}
\]

We aim to show that \( \bar{u}_t \) is a solution to the fixed point problem (2.33), (3.22). To do so, we derive now a stochastic evolution equation for \( \bar{u}_t(x) \). This will require the following two lemmas

**Lemma 3.2.** Let \( v \in C(0, T; C^2(\Omega)) \) be deterministic. Then, the process \( \theta_t := v_t \circ A_t \) solves the SPDE

\[
\begin{align*}
\frac{d}{dt} \theta_t &= \left( \partial_t v|_{A_t} - u_t \cdot \nabla \theta_t + \frac{1}{2} \sum_k (\xi^{(k)} \cdot \nabla)((\xi^{(k)} \cdot \nabla)\theta_t) \\
+ \nu \sum_k (\eta^{(k)} \cdot \nabla)((\eta^{(k)} \cdot \nabla)\theta_t) \right) dt \\
- \sum_k \xi^{(k)} \cdot \nabla \theta_t dW_t^{(k)} - \sqrt{2\nu} \sum_k \eta^{(k)} \cdot \nabla \theta_t dB_t^{(k)}.
\end{align*}
\]

\[ \tag{3.30} \]
Proof. First, the Itô form of equation (3.27) reads

\[ \text{d}_t A_t(x) + u_t(x) \cdot \nabla A_t(x) \text{d}t - \frac{1}{2} \sum_k (\xi^{(k)} \cdot \nabla)((\xi^{(k)} \cdot \nabla) A_t(x)) \text{d}t \\
- \nu \sum_k (\eta^{(k)} \cdot \nabla)((\eta^{(k)} \cdot \nabla) A_t(x)) \text{d}t \\
+ \sum_k \xi^{(k)} \cdot \nabla A_t(x) \text{d}W_t^{(k)} + \sqrt{2\nu} \sum_k \eta^{(k)} \cdot \nabla A_t(x) \text{dB}_t^{(k)} = 0. \quad (3.31) \]

Now, applying the Itô product formula, we have

\[ \text{d}\theta_t = \partial_t v_t|_{A_t} \text{d}t + dA_t \cdot \nabla v_t|_{A_t} + \frac{1}{2} d[A_t, A_t]|_t : (\nabla \otimes \nabla v_t)|_{A_t} \\
= \partial_t v_t|_{A_t} \text{d}t - (u_t \cdot \nabla A_t) \cdot \nabla v_t|_{A_t} \text{d}t - \sum_k (\xi^{(k)} \cdot \nabla A_t) \cdot \nabla v_t|_{A_t} \text{d}W_t^{(k)} \\
- \sqrt{2\nu} \sum_k (\eta^{(k)} \cdot \nabla A_t) \cdot \nabla v_t|_{A_t} \text{d}B_t^{(k)} + \frac{1}{2} d[A_t, A_t]|_t : (\nabla \otimes \nabla v_t)|_{A_t} \\
+ \frac{1}{2} \sum_k (\xi^{(k)} \cdot \nabla)((\xi^{(k)} \cdot \nabla) A_t(x)) \cdot \nabla v_t|_{A_t} \text{d}t \\
+ \nu \sum_k (\eta^{(k)} \cdot \nabla)((\eta^{(k)} \cdot \nabla) A_t(x)) \cdot \nabla v_t|_{A_t} \text{d}t. \]

Using (3.31), we compute the quadratic variation term as

\[ \frac{1}{2} d[A_t, A_t]|_t = \frac{1}{2} \sum_k (\xi^{(k)} \cdot \nabla) A_t \otimes (\xi^{(k)} \cdot \nabla) A_t \text{d}t \\
+ \nu \sum_k (\eta^{(k)} \cdot \nabla) A_t \otimes (\eta^{(k)} \cdot \nabla) A_t \text{d}t. \quad (3.32) \]

Thus, putting (3.32) together with our calculation of \( \text{d}\theta_t \), we arrive at the following equation

\[ \text{d}\theta_t = \partial_t v_t|_{A_t} \text{d}t - (u_t \cdot \nabla A_t) \cdot \nabla v_t|_{A_t} \text{d}t - \sum_k (\xi^{(k)} \cdot \nabla A_t) \cdot \nabla v_t|_{A_t} \text{d}W_t^{(k)} \\
- \sqrt{2\nu} \sum_k (\eta^{(k)} \cdot \nabla A_t) \cdot \nabla v_t|_{A_t} \text{d}B_t^{(k)} \\
+ \frac{1}{2} \sum_k (\xi^{(k)} \cdot \nabla) A_t \cdot (\nabla \otimes \nabla v_t)|_{A_t} \cdot (\xi^{(k)} \cdot \nabla) A_t \text{d}t \\
+ \nu \sum_k (\eta^{(k)} \cdot \nabla) A_t \cdot (\nabla \otimes \nabla v_t)|_{A_t} \cdot (\eta^{(k)} \cdot \nabla) A_t \text{d}t. \]
\[+ \frac{1}{2} \sum_k (\xi^{(k)} \cdot \nabla) ((\xi^{(k)} \cdot \nabla) A_t(x)) \cdot \nabla v_t|_{A_t} dt \\
+ \nu \sum_k (\eta^{(k)} \cdot \nabla) ((\eta^{(k)} \cdot \nabla) A_t(x)) \cdot \nabla v_t|_{A_t} dt.\]

Using finally the chain rule via both \((u \cdot \nabla A_t) \cdot \nabla v_t = u \cdot \nabla \theta_t\) and the identity
\[(\xi \cdot \nabla) A_t \cdot (\nabla \otimes \nabla v_t)|_{A_t} \cdot (\xi \cdot \nabla) A_t + (\xi \cdot \nabla)((\xi \cdot \nabla) A_t(x)) \cdot \nabla v_t|_{A_t} = (\xi \cdot \nabla)((\xi \cdot \nabla) \theta_t),\]
we deduce the stated evolution equation (3.30).

□

We now derive the evolution of the ‘Weber velocity’ \(w_t\), generalizing theorem 2.2 of [9] to multiplicative noise. It can also be derived as an application of Kunita’s formula (3.24) above, but we prove it here directly.

**Lemma 3.3.** Let \(v \in C(0, T; C^2(\Omega))\) and \(\theta_t := v_t \circ A_t\). The process \(w_t := (\nabla A_t(x))^T \theta_t\) solves the SPDE
\[
\begin{align*}
\frac{d}{dt} w_t + \left( L_{u_t}^T w_t - \frac{1}{2} \sum_k L_{\xi(\cdot)}^T (L_{\xi(\cdot)}^T w_t) - \nu \sum_k L_{\eta(\cdot)}^T (L_{\eta(\cdot)}^T w_t) \right) dt \\
+ \sum_k L_{\xi(\cdot)}^T w_t dW_t^{(k)} + \sqrt{2 \nu} \sum_k L_{\eta(\cdot)}^T w_t dB_t^{(k)} = 0.
\end{align*}
\]

(3.33)

**Proof.** First, the Itô form of equation (3.28) reads
\[
\begin{align*}
\frac{d}{dt} \nabla A_t(x) + L_{u_t}^T \nabla A_t(x) dt &- \frac{1}{2} \sum_k L_{\xi(\cdot)}^T (L_{\xi(\cdot)}^T \nabla A_t(x)) dt \\
- \nu \sum_k L_{\eta(\cdot)}^T (L_{\eta(\cdot)}^T \nabla A_t(x)) dt &+ \sum_k L_{\xi(\cdot)}^T \nabla A_t(x) dW_t^{(k)} + \sqrt{2 \nu} \sum_k L_{\eta(\cdot)}^T \nabla A_t(x) dB_t^{(k)} = 0.
\end{align*}
\]

(3.34)

Applying the Itô product formula, we have
\[
\frac{d}{dt} w_t = d(\nabla A_t)^T \theta_t + (\nabla A_t)^T d\theta_t + d[(\nabla A_t)^T, \theta_t].
\]

(3.35)

Using equation (3.34) and (3.30) from lemma 3.2, we compute the quadratic variation term to be
\[
\begin{align*}
\frac{d}{dt} [(\nabla A_t)^T, \theta_t] &= \sum_k L_{\xi(\cdot)}^T (\nabla A_t)^T (\xi^{(k)} \cdot \nabla \theta_t) dt + 2\nu \sum_k L_{\eta(\cdot)}^T (\nabla A_t)^T (\eta^{(k)} \cdot \nabla \theta_t) dt.
\end{align*}
\]

(3.36)
We have also that
\[
\mathrm{d}(\nabla A_t)^T \theta_t = -\mathcal{L}_{u_t}^T (\nabla A_t)^T \theta_t \mathrm{d}t + \frac{1}{2} \sum_k \mathcal{L}_{\xi(k)}^T (\mathcal{L}_{\xi(k)}^T (\nabla A_t)^T) \theta_t \mathrm{d}t
\]
\[
+ \nu \sum_k \mathcal{L}_{\eta(k)}^T (\mathcal{L}_{\eta(k)}^T (\nabla A_t)^T) \theta_t \mathrm{d}t
\]
\[
- \sum_k \mathcal{L}_{\xi(k)}^T (\nabla A_t)^T \theta_t \mathrm{d}W_t^{(k)} - \sqrt{2\nu} \sum_k \mathcal{L}_{\eta(k)}^T (\nabla A_t)^T \theta_t \mathrm{dB}_t^{(k)},
\]
\[
(\nabla A_t)^T \mathrm{d}\theta_t = -((\nabla A_t)^T u_t \cdot \nabla \theta_t \mathrm{d}t + \frac{1}{2} \sum_k ((\nabla A_t)^T (\xi(k) \cdot \nabla))((\xi(k) \cdot \nabla) \theta_t) \mathrm{d}t
\]
\[
+ \nu \sum_k ((\nabla A_t)^T (\eta(k) \cdot \nabla))((\eta(k) \cdot \nabla) \theta_t) \mathrm{d}t - \sum_k ((\nabla A_t)^T \xi(k) \cdot \nabla \theta_t) \mathrm{d}W_t^{(k)}
\]
\[
- \sqrt{2\nu} \sum_k ((\nabla A_t)^T \eta(k) \cdot \nabla \theta_t) \mathrm{d}B_t^{(k)}.
\]

For any vector field \( v \), one has the identity,
\[
\mathcal{L}_v^T u_t = (\mathcal{L}_v^T (\nabla A_t)^T) \theta_t + (\nabla A_t)^T (v \cdot \nabla) \theta_t.
\]

Consequently, the form of the noise and first drift term in (3.33) are fixed. Grouping
the remaining terms in (3.35) involving \( \xi \) and \( \eta \), using the identity (3.13) and then
performing some straightforward but tedious computations, we obtain the stated
solution equation (3.33). We remark that Kunita’s formula (3.24) can be obtained
by pairing (in \( L^2(\Omega) \)) the equation (3.33) with an arbitrary solenoidal vector field
\( v \) and integrating by parts.

Finally, let \( \tilde{u}_t(x) := (\nabla A_t(x))^T u_0(A_t(x)) \) so that \( \overline{u}_t = \mathbb{E}\{\tilde{u}_t(x)\} | F_t^{W^{(k)}} \} \). Applying lemma 3.3 to the stochastic Weber velocity \( \tilde{u}_t \), projecting onto divergence-free and averaging over the Brownian motions \( \{B^{(k)}\}_{k \in \mathbb{N}} \), we deduce that \( \overline{u}_t \) solves
the following linear SPDE
\[
\mathrm{d}_t \overline{u}_t + \mathbb{P} \left( \mathcal{L}_{u_t}^T \overline{u}_t - \frac{1}{2} \sum_k \mathcal{L}_{\xi(k)}^T (\mathcal{L}_{\xi(k)}^T \overline{u}_t) - \nu \sum_k \mathcal{L}_{\eta(k)}^T (\mathcal{L}_{\eta(k)}^T \overline{u}_t) \right) \mathrm{d}t
\]
\[
+ \sum_k \mathbb{P} \mathcal{L}_{\xi(k)}^T \overline{u}_t \mathrm{d}W_t^{(k)} = 0, \quad (3.37)
\]
with initial condition \( \overline{u}_0 = u_0 \). Clearly, one solution of (3.37) is \( u_t \) itself. Uniqueness
of the initial value problem for this type of linear stochastic system with regular
coefficients follows from the argument given in the proof of proposition 11 of [14].

Thus, we conclude that \( \overline{u}_t \equiv u_t \) for all \( t \in [0, T] \) and therefore smooth solutions \( u_t \)
of equation (2.1) solve the fixed-point problem (2.33), (2.32).

4. Discussion

In this note, we have considered two classes of stochastic models of Eulerian incompressible
fluid flow which differ in their nonlinear transport operators. These two
classes may be compared explicitly in their vector field forms, upon defining the stochastic vector field for the transport velocity written in terms of the smooth, invertible, volume-preserving flow map $X_t$ in equation (2.3) as
\[
(dX_t)X_t^{-1} := u_t dt + \sum_k \xi^{(k)} \circ dW_t^{(k)}.
\]
(4.1)
The stochastic transport operator for the energy-conserving stochastic fluid models we have treated here takes the form
\[
du_t + \mathbb{P}((dX_tX_t^{-1}) \cdot \nabla u_t) = 0.
\]
(4.2)
However, stochastic fluid models with this transport operator do not conserve the Kelvin circulation, unless the spatial gradients of their correlation eigenvectors $\xi^{(k)}$ all vanish.

In contrast, the transport operator in the stochastic fluid models we have treated here that do conserve Kelvin circulation take the form
\[
du_t + \mathbb{P}(\mathcal{L}_T^{(dX_tX_t^{-1})} u_t) = 0,
\]
(4.3)
where
\[
\mathcal{L}_T^{(dX_tX_t^{-1})} u := (\dot{X} X^{-1}) \cdot \nabla u + (\nabla (\dot{X} X^{-1}))^T \cdot u.
\]
(4.4)
Thus, the transport operators for the two classes of stochastic Euler equations treated here, in equation (4.2) which conserve energy and in equation (4.3) which conserve circulations, only differ by a single term.

Indeed, we have shown that the stochastic fluid equations in equation (4.3) are characterized by the property that circulations are conserved (pathwise in case of Euler-type models and in mean for Navier–Stokes-type) on smooth solutions. Brownian forces enter into these equations as a novel type of multiplicative noise; which involves the Lie derivative of the circulation velocity along the spatial correlation eigenvectors of the noise. This structure has geometric significance which ensures that the stochastic equations retain the Lagrangian properties of circulation, vorticity, and helicity which their deterministic counterparts possess. However, stochastic fluid models with the transport operator in equation (4.3) turn out not to conserve energy, unless the spatial gradients of their correlation eigenvectors $\xi^{(k)}$ in the cylindrical noise all vanish.

The difference between these two classes of stochastic models may appear small, especially since their transport operators exactly coincide in the deterministic case, where they each conserve both energy and Kelvin circulation. However, we have found that this difference has profound effects in the conservation properties of the stochastic fluid models treated here. Thus, the introduction of gradients into the spatial correlations of the cylindrical noise in these two classes of stochastic fluid models has introduced a sort of ‘Sophie’s choice’ between conservation of either energy, or circulation, but not both.

This comparison is summarized explicitly in the following table.

With these models in hand, we must address the following important questions: what physical systems do they represent; what insights do they yield; and how can they be exploited in practice? To address the first question, we mention
Comparison of Energy and Circulation Properties of Stochastic Fluid Models

| Euler Circulation Thm. | Euler Energy Thm. | Navier–Stokes Circulation Thm. | Navier–Stokes Energy Thm. |
|------------------------|------------------|---------------------------------|---------------------------|
| Equation (2.1) with (2.4) and (2.5) | Equation (2.1) with (2.21) and (2.22) | Equation (2.1) with (2.34) and (2.35) | Equation (2.1) with (2.43) and (2.44) |

the recent work of [11] which shows that the stochastic Euler–Poincaré equations (equation (2.1) with (2.4) and (2.5)) arise naturally upon representing the deterministic Lagrangian flow map as a composition of smooth maps with two different time scales. The first map has slowly varying time dependence. It is followed by composition with the second map which has rapidly fluctuating time dependence, with zero mean when homogenized over the rapid time scale. When dissipation is important, the corresponding Navier–Stokes–Poincaré equations (equation (2.1) with (2.34) and (2.35)) arise from similar considerations. The result of [11] shows that this stochastic model has some features in common with a deterministic regularization of the Navier–Stokes equation called the LANS-α model, which has been proposed as a model for large-scale turbulence and also preserves a certain Kelvin circulation theorem [5,27,28,38,40].

The above considerations motivate the utility of circulation-theorem preserving stochastic models as reduced descriptions of nonlinear dynamical systems which account for the advective transport effects of the small, rapid, unresolvable scales of fluid motion on the variability of computationally resolvable. See, respectively, [12, 13] for computational investigations of the Navier–Stokes–Poincaré equations in two dimensions for regions with fixed boundaries and for a 2-layer quasi-geostrophic model. See also [30] for a recent review, and see [31] for discussions of stochastic fluid models with non-stationary statistics.

On the other hand, for certain applications (depending on what observable the stochastic solution is intended to describe) it may be more important to enforce a pathwise energy balance. In this case, the models treated here which preserve the corresponding energy theorems (equation (2.1) with (2.21) and (2.22) or (2.43) and (2.44)) could be very useful. In particular, another deterministic regularization of the Navier–Stokes equation due to Leray [53] called the Leray-α model, which conserves energy in the absence of viscosity, has also been developed for simulations of turbulence and studied numerically in comparison with the LANS-α model [32, 33]. These matters will be discussed further, elsewhere.

It is only when the noise-coefficients are spatially homogeneous that these two models simultaneously preserve their respective energy and circulation theorems. Thus, as is typical in the modeling business, an application-dependent choice must be made whenever implementing these SPDEs as a practical reduced description. These issues are currently being explored and remain the subject of active and ongoing research.
Appendix A. Geometric background and notation

We discuss Kelvin’s circulation theorem from a geometric viewpoint. To begin, recall that $\text{SDiff}(\Omega)$ is the space of volume-preserving (special) diffeomorphisms, and let $X_t \in \text{SDiff}(\Omega)$ (i.e., smooth invertible flow, whose inverse is also smooth) which maps the manifold without boundaries, $\Omega \subset \mathbb{R}^d$ onto itself. Introduce the transport velocity, $u_t$, as a vector field,

$$u_t := u(x, t) : \Omega \times [0, T] \rightarrow \mathfrak{X}^{\text{sdiff}}(\mathbb{R}^d),$$

where $\mathfrak{X}(\mathbb{R}^d)$ is the space of volume-preserving vector fields defined over $\mathbb{R}^d$; so that $\nabla \cdot u_t = 0$. Next, define the corresponding circulation velocity, $u^\flat_t$, which appears in the integrand of Kelvin’s theorem,

$$u^\flat_t := u^\flat(x, t) : \Omega \times [0, T] \rightarrow \Lambda^1(\mathbb{R}^d).$$

Here $u^\flat_t$ is in the space of 1-forms $\Lambda^1(\mathbb{R}^d)$ dual to the divergence-free transport velocity vector fields, $u_t \in \mathfrak{X}(\mathbb{R}^d)$, under the $L^2$ pairing between the Lie algebra of vector fields and its dual,

$$\langle \cdot, \cdot \rangle : \Lambda^1(\mathbb{R}^d) \times \mathfrak{X}(\mathbb{R}^d) \rightarrow \mathbb{R},$$

on the domain of flow $\Omega$. Here, the familiar musical operations flat ($\flat$) and its inverse sharp ($\sharp$) essentially lower and raise vector indices, respectively, although no Riemannian metric will be needed here, because we work on $\mathbb{R}^d$. For example, the operation $\sharp : \mathfrak{X}(\mathbb{R}^d) \rightarrow \Lambda^1(\mathbb{R}^d)$ maps a vector field into a 1-form, and vice versa for $\flat$ (so that $(u^\flat_t)^\sharp = u_t$, for example). The musical notation which distinguishes between $u^\flat_t$ and $u_t$ helps one make proper mathematical sense of the operations of divergence, Lie derivative, Leray–Hodge projection, etc.

Remark A.1 (Discussion of musical isomorphisms on a Riemannian manifold). On a Riemannian manifold $M$, the musical isomorphisms are defined by the Riesz representation and inverse maps with respect to the metric. That is,

$$\sharp : T^*M \rightarrow TM \text{ as } \langle \alpha^\sharp, w \rangle = \alpha(w),$$

$$\flat : TM \rightarrow T^*M \text{ as } v_\flat(w) = \langle v, w \rangle.$$

In Riemannian geometry, the Levi–Civita connection respects the musical isomorphisms, namely for $v, w \in \mathfrak{X}(M)$ a vector field and $\alpha \in \Lambda^1(M)$ a 1-form density:

$$(\nabla_v w)_\flat = \nabla_v (w_\flat)$$

$$(\nabla_v \alpha)^\flat = \nabla_v (\alpha^\sharp).$$

Thus, the musical isomorphisms identify vector fields with 1-forms and allow them to be differentiated in the same way.
Suppose the transport and circulation velocities \( u_t \) and \( u^\flat_t \), respectively, together solve the incompressible Euler equations, written in vector form as

\[
\frac{\partial_t}{\partial t}u^\flat_t + (u_t \cdot \nabla)u^\flat_t = -\nabla p_t \tag{A.1}
\]

with scalar pressure function \( p_t \), determined by solving the Poisson equation

\[-\Delta p_t = (\nabla \otimes \nabla) : (u_t \otimes u^\flat_t).\]

The Kelvin theorem in equation (1.3) now states that any smooth Euler solution \( u_t \) has the property that for all loops \( \Gamma \subset \Omega \), the circulation integral satisfies,

\[
\oint_{X_t(\Gamma)} u^\flat_t = \oint_{X_0(\Gamma)} u^\flat_0
\]

where the time-dependent Lagrangian flow map \( X_t \) with \( X_0 = \text{id} \) is obtained by integrating the vector field

\[
\dot{X}_t = u_t(X_t) =: X^*_t u_t,
\]

where the asterix on \( X^*_t \) denotes pull back by the smooth invertible map \( X_t \). Consequently, the transport velocity vector field in the Eulerian representation is given by

\[
u_t = \dot{X}_t X_t^{-1} \in \mathfrak{X}(\mathbb{R}^d),
\]

in which the right action on the tangent vector \( \dot{X}_t \) by the inverse map \( X_t^{-1} \) (shown as concatenation from the right) translates the tangent vector along the Lagrangian path back to the identity. Thus, the Eulerian transport velocity vector field \( u_t \in \mathfrak{X}(\mathbb{R}^d) \) is right-invariant. That is, \( u_t = \dot{X}_t X_t^{-1} \) is invariant under the action of the diffeomorphisms from the right, upon transforming \( X_t \to X_t \bar{X}_t \) for any other volume-preserving diffeomorphism, \( \bar{X}_t \in \text{SDiff}(\mathbb{R}^d) \). As we shall see, right-invariance is the key to understanding the Kelvin circulation theorem from the viewpoint of Noether’s theorem.

The Kelvin theorem in (1.3) offers some insight into the geometric meaning of the Euler fluid equations. In the geometric notation introduced above, the calculation in equation (1.4) may be validated as

\[
\frac{d}{dt} \oint_{X_t(\Gamma)} u^\flat_t := \oint_{X_0(\Gamma)} \frac{d}{dt}(X^*_t u^\flat_t)
\]

\[
= \oint_{X_0(\Gamma)} X^*_t ((\partial_t + \mathcal{L}_{\dot{X}_t X_t^{-1}}) u^\flat_t)
\]

\[
= \oint_{X_t(\Gamma)} (\partial_t + \mathcal{L}_{\dot{X}_t X_t^{-1}}) u^\flat_t
\]

\[
= \oint_{X_t(\Gamma)} (\partial_t u_t + ((\dot{X}_t X_t^{-1}) \cdot \nabla) u_t + \nabla(\dot{X}_t X_t^{-1})^T \cdot u_t) \cdot d\ell = 0. \tag{A.2}
\]

In the second line of this calculation, we have used the formula [54]

\[
\frac{d}{dt}(X^*_t (u^\flat_t)) = X^*_t ((\partial_t + \mathcal{L}_{\dot{X}_t X_t^{-1}}) u^\flat_t), \tag{A.3}
\]
in which the pull-back by the flow map $X_t$ acting on the Lie derivative $\mathcal{L}_{\dot{X}_t} X_t^{-1} u_t^b$ of a 1-form $u_t^b$ with respect to the vector field $\dot{X}_t X_t^{-1}$ is defined as the time derivative of the pull-back of the 1-form $u_t^b$ by the flow map $X_t$. In the third line above in (A.2), transforming the integrand back into fixed Eulerian coordinates yields the Lie derivative itself, defined as the tangent of the pull-back, evaluated at the identity; which, as a vector expression is given by,

$$\mathcal{L}_{\dot{X}_t} X_t^{-1} u_t^b := \left[ \frac{d}{dt} (X_t^* u_t^b) \right]_{\text{id}} = ((\dot{X}_t X_t^{-1}) \cdot \nabla) u_t + \nabla(\dot{X}_t X_t^{-1})^T \cdot u_t \cdot d\ell, \quad (A.4)$$

thereby finishing the calculation.

Now, in comparing equation (1.4) with equation (A.4), we realize that the geometric meaning of the Euler fluid equations was disguised in equation (1.4), by not distinguishing between the transport velocity vector field and the circulation velocity 1-form. Of course, this distinction is unnecessary in Euclidean coordinates. However, even in Euclidean coordinates we will benefit in what follows by keeping track of this distinction. In particular, the properties of the Lie derivative will be very useful to us in what follows; and the Lie derivative of a 1-form is not the same as the Lie derivative of a vector field.

The Lie derivative of one (right-invariant, Eulerian) vector field $w$ by another one $\xi$ is defined by the following well-known formula, see, e.g., [2, 39, 41],

$$-\mathcal{L}_\xi w = -((\xi \cdot \nabla) w - (w \cdot \nabla) \xi) \cdot \nabla :=[\xi, w]=: \text{ad}_\xi w. \quad (A.5)$$

In contrast, the Lie derivative of a 1-form $v^b$ by the vector field $\xi$ is given as in the calculation (A.2) above as

$$\mathcal{L}_\xi v^b := ((\xi \cdot \nabla) v + (\nabla \xi)^T \cdot v) \cdot d\ell =: \text{ad}_\xi^* v^b. \quad (A.6)$$

In the pairing $\langle \cdot, \cdot \rangle_{L^2(\Omega)}$ with respect to the standard $L^2(\Omega)$ inner product, the operations $\text{ad}$ and $\text{ad}^*$ are dual to each other, being related by [39, 41]

$$\langle \text{ad}_\xi^* v^b, w \rangle_{L^2(\Omega)} = \langle v^b, \text{ad}_\xi w \rangle_{L^2(\Omega)} = -\langle v, \mathcal{L}_\xi w \rangle_{L^2(\Omega)}, \quad (A.7)$$

To simplify notation in what follows, we now define the adjoint operator $\mathcal{L}_\xi^T$ by the identity,

$$\langle \mathcal{L}_\xi^T v, w \rangle_{L^2(\Omega)} := \langle (\mathcal{L}_\xi^T v)^b, w \rangle_{L^2(\Omega)} = \langle \mathcal{L}_\xi v^b, w \rangle_{L^2(\Omega)} = -\langle v, \mathcal{L}_\xi w \rangle_{L^2(\Omega)}, \quad (A.8)$$

where the round brackets $(\cdot, \cdot)_{L^2(\Omega)}$ denote the usual $L^2$ integral of the dot product of vector-valued functions. Consequently, $(\mathcal{L}_\xi^T v)^b = \mathcal{L}_\xi v^b$, upon identifying corresponding terms. This relation follows due to the nondegeneracy of the $L^2(\Omega)$ pairing for a manifold without boundaries. It may also be verified by substituting (A.5) into (A.8) and integrating by parts.

Upon taking the $\sharp$ of equation (A.1) to transform it from 1-forms to vector fields and applying the Leray–Hodge projection $\mathbb{P}$, it becomes $\mathcal{L} \mathbb{P} = 0$. 

$$du_t + \mathbb{P}(\mathcal{L}_{u_t} u_t) dt = 0.$$
Thus, the corresponding equation for the vector field \( u_t = u_t \cdot \nabla \) can be expressed as

\[
du_t + \mathbb{P}(\text{ad}^\dagger_{u_t} u_t) dt = 0,
\]

where the binary operation among vector fields \( \text{ad}^\dagger : \mathfrak{X} \times \mathfrak{X} \rightarrow \mathfrak{X} \) is defined for vector fields \( \xi \) and \( v \) by

\[
\text{ad}^\dagger \xi v := (\text{ad}^\ast \xi v) = (L_\xi v^\flat) = : L_\xi^T v.
\]

Having identified \( \text{ad}^\dagger \xi v = L_\xi^T v \), from equations (A.7), (A.8) and (A.9), we see that the musical notations sharp (\( \# \)) and flat (\( \flat \)) can now be replaced by the simpler \( L_\xi^T \) notation. Namely, in what follows, we will distinguish notationally between components of Lie-derivative operations on vector fields and 1-forms as,

\[
-(L_\xi v)^\# := \xi \cdot \nabla v + \nabla \xi \cdot v.
\]

This notation distinguishes between divergence free vector fields and their \( L^2 \)-dual 1-forms only by whether the action of vector fields \( \xi \) on them appears as \( L_\xi \) or \( L_\xi^T \).

We note that the operation \( L_\xi^T \) is denoted as \( \mathcal{B}(\xi, \cdot) \) in [2], as may be identified in the following relation,

\[
(B(\xi, v), w)_{L^2(\Omega)} = (v, \nabla \times \nabla \xi \cdot v)_{L^2(\Omega)}.
\]

The operator \( \mathcal{B} \) in [2] is distinct from \( B(w, v) := \mathbb{P}(w \cdot \nabla v) \) (see e.g. [8]) introduced for equation (2.21).

Remark A.2 (Commutator in three-dimensional Euclidean space). As we see above in equation (A.10), the commutator of two (right-invariant) vector fields is (minus) their Lie derivative. The commutator of divergence-free vector fields in a three-dimensional Euclidean space \( \mathbb{R}^3 \) is given by the formula

\[
-(L_\xi w) := [\xi, w] = \text{curl}(\xi \times w),
\]

where \( \xi \times w \) is the cross product. Hence, we may rewrite the relations in equation (A.8) in this notation as

\[
(L_\xi^T v, w)_{L^2(\Omega)} = (v, [\xi, w])_{L^2(\Omega)} = (v, \text{curl}(\xi \times w))_{L^2(\Omega)} = \text{curl}(\xi \times w, v)_{L^2(\Omega)}.
\]

Thus, we find, in ordinary vector notation, \( L_\xi^T v = -\xi \times \text{curl} v \), modulo a gradient term, since \( \nabla \cdot w = 0 \).
Appendix B. Variational Principle for the Stochastic Euler–Poincaré equations

In this appendix, we treat only the formal aspects of stochastic variational principles in infinite dimensions, for the purpose of modeling time-dependent spatial correlations. As discussed in remark 2.4, some of the fundamental questions in analysis for the stochastic 3D Euler–Poincaré fluid model have been answered in [14], who proved local in time existence, uniqueness, and well posedness of their solutions in regular spaces, as well as a Beale–Kato–Majda blow-up criterion for these equations. These are precisely the same analytical properties as for the deterministic 3D Euler fluid equations. Thus, in this case, introducing stochasticity that preserved the geometric properties of the Euler fluid equations also preserved their analytical properties. The corresponding questions still remain open for the other stochastic fluid models discussed here.

Brief history of variational derivations of the Navier–Stokes equations from stochastic equations

The derivation of Navier–Stokes equations in the context of stochastic processes has a long and well-known history. See, e.g., Constantin and Iyer [9], Eyink [20], and references therein. Previous specifically variational treatments of fluid equations generally started from the famous remark by Arnold [1966] about Euler’s equations for the incompressible flow of an ideal fluid being geodesic for kinetic energy given by the $L^2$ norm of fluid velocity, and they have mainly treated Itô noise in this variational context. For more discussion of variational derivations of fluid equations and their relation to the stochastic processes, one should consult original sources such as, in chronological order, Inoue and Funaki [43], Rapoport [58, 59], Gomes [35], Cipriano and Cruzeiro [6], Constantin and Iyer [9], Eyink [18], Gliklikh [34], Arnaudon, Chen and Cruzeiro [1], Eyink and Drivas [21]. For additional information, review and background references for random perturbations of PDEs and fluid dynamic models, viewed from complementary viewpoints to the present paper, see also Flandoli et al. [23, 26]. In particular, the latter paper studies the interesting possibility that adding stochasticity would have a regularizing effect on fluid equations which might otherwise be ill-posed. However, it is not our intention to make another variational derivation of the Navier–Stokes equations in the present work, as we explain below.

Appendix B.1. The stochastic Hamilton–Pontryagin variational principle [31]

We proceed formally here and below to derive the stochastic Euler–Poincaré equations in (1.5), by considering the reduced stochastic Hamilton–Pontryagin (RSHP) principle in which the Lagrangian path in equation (2.3) in theorem 2.1 is written in Eulerian coordinates and imposed as a constraint on variations in the Eulerian
representation of Hamilton’s principle, as

$$\delta \int_0^T [l(u_t)dt + \langle m, (dX_t)X_t^{-1} - u_t dt - \sum_k \xi^{(k)} \circ dW_t^{(k)} \rangle] \bigg|_{L^2(\Omega)} = 0, \quad (B.1)$$

with respect to variations $\delta u_t, \delta X_t, \delta m$, for the Lagrangian functional $l(u_t)$. This is the reduced stochastic Hamilton–Pontryagin (RSHP) principle found in [31]. Its Eulerian stationarity conditions are

$$\delta m : \quad (dX_t)X_t^{-1} = u_t dt + \sum_k \xi^{(k)} \circ dW_t^{(k)},$$
$$\delta u_t : \quad \frac{\delta l}{\delta u_t} = m,$$
$$\delta X_t : \quad dm + ad^*_{(dX_t)X_t^{-1}} m = 0,$$

where we have applied the formula for integration by parts for a Stratonovich stochastic process [57] in computing the dynamics of the Lagrange multiplier, $m$. In this computation, we have also used the relation,

$$\delta((dX_t)X_t^{-1}) = dw - ad_{(dX_t)X_t^{-1}} w,$$

and dropped the endpoint term $\langle m, w \rangle|_0^T$, since the variation $\delta X$ vanishes at the endpoints of interval $[0, T]$.

In the Euler fluid case, the Lagrangian is the fluid kinetic energy

$$l(u_t) = \frac{1}{2} ||u_t||^2_{L^2(\Omega)}$$

and its variation with respect to the velocity vector field is given by the circulation 1-form,

$$m = \frac{\delta l}{\delta u_t} = u_t^\flat.$$

Now, taking the $\xi$ of the variational equation for $m$ above and using the divergence free property of the vector field $(dX_t)X_t^{-1}$ in the pairing, yields the velocity vector-field equation,

$$0 = d(\mathbb{P}m^\sharp) + \mathbb{P}(\mathcal{L}_{(dX_t)X_t^{-1}} m)^\sharp = du_t + \mathbb{P}(\mathcal{L}_{(dX_t)X_t^{-1}} u_t).$$

Thus, for the Euler case, the stochastic RSHP principle in (B.1) yields the stochastic Euler–Poincaré motion equation in (1.5).

The Eulerian vector field $(dX_t)X_t^{-1} \in \mathfrak{X}(\mathbb{R}^d)$ is invariant under the action of the diffeomorphisms from the right, given by $X_t \rightarrow X_t \bar{X}$ for any fixed volume-preserving diffeomorphism $\bar{X} \in SDiff(\mathbb{R}^d)$. Since the motion of a Lagrangian trajectory is given by applying $X_t$ to an initial condition $x_0$, this symmetry simply corresponds to well-known invariance of the Eulerian fluid velocity vector field $u_t$ under relabeling of the Lagrangian coordinates as $x_0 \rightarrow \bar{X}x_0$. 


Appendix B.2. Noether’s theorem and preservation of Kelvin circulation

The Eulerian vector field \((dX_t) X_t^{-1} \in \mathfrak{x}(\mathbb{R}^d)\) is invariant under the action of the diffeomorphisms from the right, given by \(X_t \rightarrow X_t \bar{X}\) for any fixed volume-preserving diffeomorphism \(\bar{X} \in \text{SDiff}(\mathbb{R}^d)\). Since the motion of a Lagrangian trajectory is given by applying \(X_t\) to an initial condition \(x_0\), this symmetry corresponds to well-known invariance of the Eulerian fluid velocity vector field \(u_t\) under relabeling of the Lagrangian coordinates as \(x_0 \rightarrow \bar{X}x_0\). As we shall see below, this symmetry of the Eulerian velocity vector field under relabeling of Lagrangian particles by smooth invertible maps leads via Noether’s theorem to conservation of circulation around each fluid loop and also to conservation of an integrated topological quantity known as helicity. See Chapter 4 of [42] for a review of Noether’s theorem relating symmetries and conservation laws.

The endpoint term arising from integration by parts in the RSHP variational principle is \(\langle m, w \rangle\), as shown above. Vanishing of the endpoint term leads to the variational equations of motion. However, according to Noether’s theorem, if \(\delta S = 0\) due to invariance of the Lagrangian under a Lie symmetry transformation, then the endpoint term will keep its value under the evolution governed by variational equations. In the present case, the right-invariant vector field \(w\) generates an arbitrary time-independent diffeomorphism of the reference flow domain, under which the Lagrangian is invariant, since the Eulerian representation is invariant under a volume-preserving diffeomorphism of the Lagrangian parcel labels.

In the Euler fluid case, \(m\) is a 1-form density and the quantity \(m/D\) is a 1-form, although we can ignore the difference, since \(D = 1\) results as the Jacobian for the flow map generated by a divergence-free vector field. Thus, we can regard \(m = u_t^\flat = u \cdot dx\) as simply a 1-form, which is evolving by coadjoint action on it by the diffeomorphism \(X_t\), so that it satisfies

\[
X_t^*(dm + \text{ad}_{(dX_t)X_t^{-1}}^* m) = \frac{d}{dt}(X_t^* m) = -X_t^*(dp),
\]

where \(X_t^*\) is the pullback of the Lagrangian flow and we have introduced \(-dp = -\nabla p \cdot dx\) to account for incompressibility of \(m^\sharp\). The previous equation implies that the integral of the 1-form \(m\) around any loop that moves with the flow is constant, as a result of its RSHP equation of motion. Thus, by Noether’s theorem, invariance of the Eulerian form of the fluid Lagrangian under fluid particle relabeling implies preservation of Kelvin’s circulation integral.

**Remark B.1 (Conservation of helicity).** The previous equation is equivalent to the Eulerian expression,

\[
dm = -\mathcal{L}_{(dX_t)X_t^{-1}} m - dp.
\]

Consequently, the stochastic evolution of the helicity density (a 3-form) is given by

\[
d(m \wedge dm) = -(\mathcal{L}_{(dX_t)X_t^{-1}} m + dp) \wedge dm) - m \wedge (\mathcal{L}_{(dX_t)X_t^{-1}} dm)
\]

\[
= -\mathcal{L}_{(dX_t)X_t^{-1}} (m \wedge dm) - d(p \ dm)
\]

\[
= -\text{div}((dX_t)X_t^{-1})(u \cdot \text{curl}u) + p\text{curl}u)d^3x.
\]
For homogeneous boundary conditions, this implies the conservation of the helicity integral,

\[ \frac{d}{dt} \int_{\Omega} (m \wedge \mathbf{d}m) = \int_{\Omega} \mathbf{u} \cdot \text{curl} \mathbf{u} \, d^3x = 0, \]

which is interpreted as the conservation of the average self-linking number of vorticity field lines, [3, 65].

Appendix B.3. Passing to the Lie–Poisson Hamiltonian formulation

The Noether quantity also plays an important geometric role on the Hamiltonian side. The reduced Legendre transformation in the Eulerian representation is given by, cf. equation (B.1),

\[
\begin{align*}
    h(m) &= \langle m, (\mathbf{d}X_t^t)X^{-1}_t \rangle_{L^2} \\
    &\quad - \left[ l(u_t) \, dt + \left\langle m, (\mathbf{d}X_t^t)X^{-1}_t - u_t \, dt - \sum_k \xi^{(k)} \circ dW_t^{(k)} \right\rangle_{L^2} \right] \\
    &= \langle m, u_t \, dt + \sum_k \xi^{(k)} \circ dW_t^{(k)} \rangle_{L^2} - l(u_t) \, dt \\
    &= \frac{1}{2} \langle m, m^2 \rangle_{L^2} \, dt + \langle m, \sum_k \xi^{(k)} \rangle_{L^2} \circ dW_t^{(k)},
\end{align*}
\]

where we have used \( l(u_t) = 1/2 \langle u^t, u \rangle_{L^2} = 1/2 \langle m, m^2 \rangle_{L^2} \) and the symmetry of the pairing \( \langle \cdot, \cdot \rangle_{L^2} \) to simplify and regroup terms in the final step of deriving the reduced Hamiltonian, \( h(m) \). We note that the stochastic part of the Hamiltonian \( h(m) \) in (B.2) couples the noise to the momentum map by \( L^2 \) pairing. The variational derivative of \( h(m) \) with respect to \( m \) returns the original stochastic Eulerian vector field,

\[ \frac{\delta h}{\delta m} = m^2 \, dt + \sum_k \xi^{(k)} \circ dW_t^{(k)} = (\mathbf{d}X_t^t)X^{-1}_t. \]

Finally, we may rearrange the Euler fluid motion equation into the Lie–Poisson Hamiltonian form [39]

\[ \mathbf{d}m = - \text{ad}_{\delta h/\delta m}^* m = \{m, h(m)\}, \]

in which the stochastic Hamiltonian is given above in the last line in equation (B.2), and the Lie–Poisson bracket for functionals \( f \) and \( h \) is defined by

\[
\begin{align*}
    df(m) &= \{f(m), h(m)\} := -\left\langle m, \text{ad}_{\delta h/\delta m} \frac{\delta f}{\delta m} \right\rangle_{L^2} = -\left\langle m, \left[ \frac{\delta h}{\delta m}, \frac{\delta f}{\delta m} \right] \right\rangle_{L^2}. \\
    \end{align*}
\]

This Lie–Poisson bracket satisfies the Jacobi identity, because it is a linear functional of the Lie bracket for the Lie algebra of divergence-free vector fields, which is known to satisfy the Jacobi identity.
Nonconservation of the deterministic energy under this Hamiltonian dynamics can be checked easily by setting $f(m) = \frac{1}{2} \langle m, m \rangle = \frac{1}{2} \| u_t \|^2_{L^2}$ in the equation (B.3) and denoting $\Xi := \sum_k \xi^{(k)} \circ dW_t^{(k)}$, to find,

$$
\frac{1}{2} \langle m, m \rangle_{L^2(\Omega)} \, \xi^{(k)} = -\langle \text{ad}^*_\xi \circ [dX_t^{-1}], m \rangle_{L^2(\Omega)} = -\langle \nabla \xi \circ (\text{curl} \, u), \Xi \rangle_{L^2(\Omega)} = -\int_\Omega \sum_k \xi^{(k)} \cdot \nabla \, u \, \Xi \, d^3x,
$$

where we have used the divergence-free property twice in the last line, when integrating by parts. This result is the Stratonovich version of equation (2.40) when viscosity $\nu$ is absent. Namely, the original deterministic fluid kinetic energy is not conserved under the evolution of the circulation conserving stochastic fluid model, unless the spatial gradients of the correlation eigenvectors $\xi^{(k)}$ vanish.

**Remark B.2 (Purely stochastic passive 1-form transport).** If we simply drop the fluid kinetic energy in the total Hamiltonian $h(m)$ in equation (B.2), then only the stochastic part would remain. Consequently, the Lie–Poisson bracket in (B.3) would produce a linear passive 1-form transport equation given by

$$
dm + \text{ad}^*_\xi \circ dW_t^{(k)} \, m = -dp,
$$

where $p$ is determined by requiring that the gauge $\nabla \cdot A_t = 0$ be preserved. In our other notation, the above equation can be written for a 1-form $A_t$ as

$$
dA_t - \sum_k \mathbb{P} (\mathcal{L}_{\xi^{(k)}} A_t) \circ dW_t^{(k)} = 0,
$$

where $\mathcal{L}_{\xi^{(k)}} A_t = \xi^{(k)} \cdot \nabla A_t + (\nabla \xi^{(k)})^T \cdot A_t$, in vector notation. Equation (B.4) is the dual problem to the passive Lie-transport equation for the vector field $B_t = \text{curl} A_t$,

$$
dB_t + \sum_k \mathcal{L}_{\xi^{(k)}} B_t \circ dW_t^{(k)} = 0,
$$

where $\mathcal{L}_{\xi^{(k)}} B_t = (\xi^{(k)} \cdot \nabla) B_t - (B_t \cdot \nabla) \xi^{(k)}$ in vector notation. Note that, since $\nabla \cdot A_t = 0$, the field $A_t$ can be recovered uniquely from $B_t$ via the Biot-Savart law $A_t = (-\Delta)^{-1} \text{curl}(B_t)$. In parallel with remark B.1, for this linear stochastic transport problem, the magnetic helicity $\langle A_t, B_t \rangle_{L^2(\Omega)}$ is conserved pathwise.

The equation (B.5) is known as the Kazantsev–Kraichnan model of kinematic dynamo, in which $B_t$ represents a transported magnetic field by a white-in-time Gaussian advecting velocity which is typically assumed to be spatially rough [45, 47, 48]. Not unexpectedly, when the noise correlates $\{\xi^{(k)}\}_{k \in \mathbb{N}}$ are smooth, the Kelvin theorem for equation (B.4) preserves the circulation around closed loops which are transported along stochastic Lagrangian paths in the Stratonovich sense. In this setting, the circulation integral represents the gauge-invariant magnetic flux.
and the conservation law corresponds to Alfvén’s theorem. The stochastically propagating closed loops must each retain its linkage number; since diffeomorphisms cannot change the topology of a curve embedded in the flow, even if the flow has a stochastic time dependence. This may fail to be true in the Kazantsev–Kraichnan model in which the fields \( \{ \xi^{(k)} \}_{k \in \mathbb{N}} \) are assumed to be only Hölder continuous \( C^\alpha(\Omega) \) with exponent \( \alpha \in (0, 1) \). In this case, Lagrangian trajectories in fixed realizations of the advecting Gaussian velocity may become non-unique and the phenomenon of spontaneous stochasticity \([4, 15]\) must be accounted for when discussing Lagrangian transport properties, see \([16–18]\).

Acknowledgements

We are enormously grateful for many helpful and inspiring discussions with P. Constantin, A.B. Cruzeiro, D. Crisan, G.L. Eyink, F. Flandoli, J.-M. Leahy, E. Mémin, H.Q. Nguyen, T.S. Ratiu, V. Resseguier and S. Takao. Research of TD is partially supported by NSF-DMS grant 1703997. DDH is grateful for partial support by the EPSRC Standard Grant EP/N023781/1.

References

1. M. Arnaudon, X. Chen and A. B. Cruzeiro. Stochastic Euler–Poincaré reduction. *J. Math. Physics* **55** (2014), 081507.
2. V. I. Arnold and B. A. Khesin. *Topological methods in hydrodynamics* (New York: Springer, 1998).
3. V. I. Arnold and B. A. Khesin. *Topological methods in hydrodynamics*, vol. 125 (Springer Science & Business Media, 1999).
4. D. Bernard, K. Gawedzki and A. Kupiainen. Slow modes in passive advection. *J. Stat, Phys.* **90** (1998), 519–569.
5. S. Chen, C. Foias, D. D. Holm, E. J. Olson, E. S. Titi and S. Wynne. The Camassa-Holm equations as a closure model for turbulent channel and pipe flows. *Phys. Rev. Lett.* **81** (1998), 5338–5341.
6. F. Cipriano and A. B. Cruzeiro. Navier–Stokes equation and diffusions on the group of homeomorphisms of the torus. *Comm. Math. Phys.* **275** (2007), 255–269.
7. P. Constantin. An Euler–Lagrangian approach for incompressible fluids: local theory. *J. Amer. Math. Soc.* **14** (2001), 263–27.
8. P. Constantin and C. Foias. *Navier-stokes equations* (University of Chicago Press, 1988).
9. P. Constantin and G. Iyer. A stochastic Lagrangian representation of the three-dimensional incompressible Navier–Stokes equations. *Commun. Pure Appl. Math* **61** (2008), 330–345.
10. P. Constantin and G. Iyer. A stochastic-Lagrangian approach to the Navier–Stokes equations in domains with boundary. *The Annals of Applied Probability* **21** (2011), 1466–1492.
11. C. J. Cotter, G. A. Gottwald and D. D. Holm. Stochastic partial differential fluid equations as a diffusive limit of deterministic Lagrangian multi-time dynamics. *Proc. R. Soc. A* **473** (2017), 20170388.
12. C. Cotter, D. Crisan, D. D. Holm, W. Pan and I. Shevchenko. Numerically modelling stochastic Lie transport in fluid dynamics. arXiv preprint arXiv:1801.09729 (2018).
13. C. Cotter, D. Crisan, D. D. Holm, W. Pan and I. Shevchenko. Modelling uncertainty using circulation-preserving stochastic transport noise in a 2-layer quasi-geostrophic model. arXiv preprint arXiv:1802.05711 (2018).
14. D. Crisan, F. Flandoli and D. D. Holm. Solution properties of a 3D stochastic Euler fluid equation. *J Nonlinear Sci* **29** (2018), 813–870.
15. T. D. Drivas. Anomalous Dissipation, Spontaneous Stochasticity & Onsager’s Conjecture. Diss. Johns Hopkins University (2017).
16 T. D. Drivas and G. L. Eyink. A Lagrangian fluctuation–dissipation relation for scalar turbulence. Part I. Flows with no bounding walls. *J. Fluid Mech.* **829** (2017), 153–189.
17 T. D. Drivas and G. L. Eyink. A Lagrangian fluctuation–dissipation relation for scalar turbulence. Part II. Wall-bounded flows. *J. Fluid Mech.* **829** (2017), 236–279.
18 G. L. Eyink. Turbulent diffusion of lines and circulations. *Phys. Lett A* **368** (2007), 486–490.
19 G. L. Eyink. Stochastic line motion and stochastic flux conservation for nonideal hydro-magnetic models. *J. Math. Phys.* **50** (2009), 083102.
20 G. L. Eyink. Stochastic least-action principle for the incompressible Navier–Stokes equation. *Phys. D: Nonlinear Phenomena* **239** (2010), 1236–1240.
21 G. L. Eyink and T. D. Drivas. Spontaneous stochasticity and anomalous dissipation for Burgers equation. *J. Stat. Phys.* **158** (2015), 386–432.
22 F. Flandoli. *Random Perturbation of PDEs and Fluid Dynamic Models: École d’été de Probabilités de Saint-Flour XL 2010*, vol. 2015 (Springer Science & Business Media, 2011).
23 F. Flandoli. *Random Perturbation of PDEs and Fluid Dynamic Models, Saint Flour summer school lectures 2010*, Lecture Notes in Mathematics n. 2015 (Berlin: Springer, 2011).
24 F. Flandoli and D. Gatarek. Martingale and stationary solutions for stochastic Navier–Stokes equations. *Probab. Theory Related Fields* **102** (1995), 367–391.
25 F. Flandoli and D. Luo. Euler–Lagrangian approach to 3D stochastic Euler equations. arXiv preprint arXiv:1803.05319 (2018).
26 F. Flandoli, M. Maurelli and M. Neklyudov. Noise prevents infinite stretching of the passive field in a stochastic vector advection equation. *J. Math. Fluid Mech.* **16** (2014), 805–822.
27 C. Foias, D. D. Holm and E. S. Titi. The Navier–Stokes–alpha model of fluid turbulence. *Phys. D* **152–153** (2001), 505–519.
28 C. Foias, D. D. Holm and E. S. Titi. The three dimensional viscous Camassa-Holm equations, and their relation to the Navier-Stokes equations and turbulence theory. *J. Dyn. Diff. Equats.* **14** (2002), 1–35.
29 A. Friedman. *Stochastic Differential Equations and Applications* (Mineola, NY: Dover, 2006).
30 F. Gay–Balmaz and D. D. Holm. Predicting Uncertainty in Geometric Fluid Mechanics. arXiv preprint arXiv:1806.10470 (2018).
31 F. Gay–Balmaz and D. D. Holm. Stochastic geometric models with non-stationary spatial correlations in Lagrangian fluid flows. *J Nonlinear Sci* **28** (2018), 873–904.
32 B. J. Geurts and D. D. Holm. Alpha-modeling strategy for LES of turbulent mixing. In *Turbulent Flow Computation* (ed. D. Drikakis and B. G. Geurts), pp. 237–278 (London: Kluwer, 2002).
33 B. J. Geurts and D. D. Holm. Regularization modeling for large-eddy simulation. *Phys. Fluids* **15** (2003), L13–L16.
34 Y. E. Gliklikh. Solutions of Burgers, Reynolds, and Navier–Stokes equations via stochastic perturbations of inviscid flows. *J. Nonlin. Math. Phys.* **17** (sup1 (2010), 15–29.
35 D. A. Gomes. A variational formulation for the Navier–Stokes equation. *Comm. Math. Phys.* **257** (2005), 227–234.
36 H. Helmholtz. Über Integrale der hydrodynamischen Gleichungen welche den Wirbelbewegungen entsprechen. *Crelles J.* **55** (1858), 25–55.
37 D. D. Holm. Variational principles for stochastic fluid dynamics. *Proc. R. Soc. A* **471** (2015), 20140963.
38 D. D. Holm, J. E. Marsden and T. S. Ratiu. Euler–Poincaré models of ideal fluids with nonlinear dispersion. *Phys. Rev. Lett.* **349** (1998), 4173–4177.
39 D. D. Holm, J. E. Marsden and T. S. Ratiu. The Euler–Poincaré equations and semidirect products with applications to continuum theories. *Adv. Math.* **137** (1998), 1–81.
40 D. D. Holm, C. Jeffery, S. Kurien, D. Livescu, M. A. Taylor and B. A. Wingate. The LANS-α model for computing turbulence. *Los Alamos Sci.* **25** (2005), 152–171.
41 D. D. Holm, T. Schmah and C. Stoica. *Geometric mechanics and symmetry: from finite to infinite dimensions*, vol. 12 (Oxford University Press, 2009).
42 D. D. Holm, T. Schmah and C. Stoica. *Geometric Mechanics and Symmetry: from Finite to Infinite Dimensions*, vol. 12 (Oxford University Press, 2009).
43 A. Inoue and T. Funaki. On a new derivation of the Navier–Stokes equation. *Commun. Math. Phys.* **65** (1979), 83–90.
44 G. Iyer. A stochastic perturbation of inviscid flows. *Commun. Math. Phys.* **266** (2006), 631–645.
45 A. P. Kazantsev. Enhancement of a magnetic field by a conducting fluid. *Sov. Phys. JETP* **26** (1968), 1031–1034.
46 T. Kelvin. On vortex motion. *Trans. Roy. Soc. Edinb.* **25** (1869), 217–260.
47 R. H. Kraichnan. Small-scale structure of a scalar field convected by turbulence. *Phys. Fluids* **11** (1968), 945–953.
48 R. H. Kraichnan and S. Nagarajan. Growth of turbulent magnetic fields. *Phys. Fluids* **10** (1967), 859–870.
49 N. V. Krylov and B. L. Rozovskii. Stochastic evolution equations. Itogi Nauki i Tekhniki. Seriya “Sovremennye Problemy Matematiki. Novsheie Dostizheniya” **14** (1979), 71–146.
50 H. Kunita. *Some extensions of Ito’s formula. Séminaire de Probabilités XV 1979/80*, pp. 118–141 (Berlin, Heidelberg: Springer, 1981).
51 H. Kunita. *Stochastic Flows and Stochastic Differential Equations, Cambridge Studies in Advanced Mathematics*, vol. 24 (Cambridge, UK: Cambridge Univ. Press, 1997).
52 Y. Le Jan and S. Watanabe. *Stochastic flows of diffeomorphisms. North-Holland Mathematical Library*, vol. 32, pp. 307–332 (Elsevier, 1984).
53 J. Leray. Sur les movements d’un fluide visqueux remplissant l’espace. *Acta Math.* **63** (1934), 193.
54 J. E. Marsden and T. J. Hughes. Mathematical Foundations of Elasticity. Courier Corporation (1994).
55 E. Mémin. Fluid flow dynamics under location uncertainty. *Geophys. Astrophys. Fluid Dyn.* **108** (2014), 119–146.
56 R. Mikulevicius and B. L. Rozovskii. Global $L^2$-solutions of stochastic Navier–Stokes equations. *Annals of Probab.* **33** (2005), 137–176.
57 P. E. Protter. Stochastic differential equations. In *Stochastic integration and differential equations*, pp. 249–361 (Berlin, Heidelberg: Springer, 2005).
58 D. L. Rapoport. Stochastic differential geometry and the random integration of the Navier–Stokes equations and the kinematic dynamo problem on smooth compact manifolds and Euclidean space. *Hadronic J.* **23** (2000), 637–675.
59 D. L. Rapoport. On the geometry of the random representations for viscous fluids and a remarkable pure noise representation. *Rep. Math. Phys.* **50** (2002), 211–250.
60 E. Resseguier, E. Mémin and B. Chapron. Geophysical flows under location uncertainty, Part I: Random transport and general models. *Geophys. Astrophys. Fluid Dyn.* **111** (2017a), 149–176.
61 E. Resseguier, E. Mémin and B. Chapron. Geophysical flows under location uncertainty, Part II: Quasigeostrophic models and efficient ensemble spreading. Geophys. Astrophys. Fluid Dyn. **111** (2017b), 177–208.
62 E. Resseguier, E. Mémin and B. Chapron. Geophysical flows under location uncertainty, Part III: SQG and frontal dynamics under strong turbulence. *Geophys. Astrophys. Fluid Dyn.* **111** (2017c), 209–227.
63 F. Rezakhanlou. Regular flows for diffusions with rough drifts. arXiv preprint arXiv:1405.5856 (2014).
64 F. Rezakhanlou. *Stochastically symplectic maps and their applications to the Navier–Stokes equation*. Annales de l’Institut Henri Poincare (C) Non Linear Analysis. vol. 33, No.1 (Masson: Elsevier, 2016).
65 D. Serre. Helicity and other conservation laws in perfect fluid motion. *Comptes Rendus Mécanique* **346** (2018), 175–183.
66 F. Shizan and D. Luo. Constantin and Iyer’s Representation Formula for the Navier–Stokes Equations on Manifolds. *Potential Anal.* **48** (2018), 181–206.