



Gamma limits and U -statistics on the Poisson space

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Abstract. Using Stein’s method and the Malliavin calculus of variations, we derive explicit estimates for the Gamma approximation of functionals of a Poisson measure. In particular, conditions are presented under which the distribution of a sequence of multiple Wiener-Itô stochastic integrals with respect to a compensated Poisson measure converges to a Gamma distribution. As an illustration, we present a quantitative version and a non-central extension of a classical theorem by de Jong in the case of degenerate U -statistics of order two. Several multidimensional extensions, in particular allowing for mixed or hybrid limit theorems, are also provided.

1. Introduction

The use of the Malliavin calculus of variations in order to deduce limit theorems for non-linear functionals of random measures has recently become a relevant direction of research, one reason for that being the many successful applications in geometric probability or stochastic geometry. Apart from a few exceptions, most contributions to this topic fall into the two categories of normal and Poisson approximations; see [Decreusefond et al. \(2011\)](#); [Lachièze-Rey and Peccati \(2013a,b\)](#); [Last](#)

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et al. (2012); Peccati et al. (2010); Peccati and Zheng (2010); Reitzner and Schulte (2012+); Viquez (2011) for distinguished examples of the former class, mostly based on the use of the Stein's method (cf. Nourdin and Peccati (2012)); see Bourguin and Peccati (2012); Peccati (2011); Schulte and Thäle (2012) for references based on the combination of Malliavin calculus and of the Chen-Stein method for Poisson approximations. We also refer to Eden and Viquez (2012) for recent extensions to general absolutely continuous distributions having support equal to the real line.

The aim of the present paper is to provide the first array of results concerning limit theorems on the Poisson space, where the limit distribution is absolutely continuous and has support contained in a proper subset of \mathbb{R} . More precisely, we are interested in probabilistic approximations where the limiting random variable has a centred Gamma distribution $\bar{\Gamma}_\nu$ with parameter $\nu > 0$. We say that a random variable $G(\nu)$ has distribution $\bar{\Gamma}_\nu$ if $G(\nu) \stackrel{d}{=} 2F(\nu/2) - \nu$, where $F(\nu/2)$ has a usual Gamma distribution with mean and variance both equal to $\nu/2$ (here and throughout $\stackrel{d}{=}$ stands for equality in distribution). If $\nu \geq 1$ is an integer, then $\bar{\Gamma}_\nu$ reduces to the centred χ^2 -distribution with ν degrees of freedom. We remark that the support of $\bar{\Gamma}_\nu$ is given by the half-line $[-\nu, +\infty)$, and that the first four moments of $\bar{\Gamma}_\nu$ are 0, 2ν , 8ν and $12\nu^2 + 48\nu$, respectively. We will often meet these expressions in the discussion to follow.

Our main contribution is the general estimate stated in Theorem 2.1, which involves Malliavin operators and is obtained by means of Stein's method, allowing one to measure the distance between the law of a given Poisson functional and $\bar{\Gamma}_\nu$. This estimate is applied to deduce explicit sufficient conditions for Gamma limit theorems involving sequences of multiple Wiener-Itô stochastic integrals. Our analysis is significantly inspired by Nourdin and Peccati (2009a,b), where the problem addressed in the present paper was first studied in the framework of non-linear functionals of general Gaussian fields. However, due to the combinatorial complications one has to face when dealing with point measures, our paper contains a number of new subtle computations related to the explicit estimation of Malliavin operators on configuration spaces. One specific problem we will have to deal with is that the solution of the Stein's equation associated with the law of $G(\nu)$ is not differentiable at $x = -\nu$. Thus, in order to obtain bounds that are well-suited for our applications (which may involve random variables possibly taking values in $(-\infty, -\nu)$), we will have to combine techniques recently introduced by Schulte (2012) with classical isometric formulae borrowed from the standard reference Privault (2009); see Proposition 2.3 below. One should note that, in view of the exact chain rules that are available on a Gaussian space, the non-differentiability of the Stein solution in one point is immaterial when studying the Gamma approximation of smooth functionals of a Gaussian field; see again Nourdin and Peccati (2009a,b).

As an illustration, we will include some applications to non-central limit theorems for sequences of degenerate (in the sense of Hoeffding) U -statistics. Our findings generalize several classic result in the field; cf. Bhattacharya and Ghosh (1992); Jammalamadaka and Janson (1986). In particular, we derive a quantitative and a non-central version of a famous theorem by de Jong (1987, 1990). Our analysis also contains a quantitative version of a non-central result recently discussed by Reitzner and Schulte (2012+, Section 5.1).

Finally, to demonstrate the flexibility and scope of our approach, we will show that our analysis can naturally be extended to a multidimensional framework. We

will not only obtain multidimensional Gamma limit theorems, but also *mixed* or *hybrid* results, where the multidimensional limit distribution is composed both of Gamma and of normal or Poisson components. This kind of limit theorems heavily relies on our use of Malliavin operators. We are not aware of any other available technique allowing one to deduce general mixed limit results, such as the ones deduced in the present paper. We shall see that our findings are a refinement of the ‘Portmanteau inequalities’ recently obtained by Bourguin and Peccati in [Bourguin and Peccati \(2012\)](#). In this respect, we stress that our results will implicitly yield a collection of sufficient conditions in order to have that two sequences of Poisson functionals are asymptotically independent. This provides a new contribution to the difficult and mostly open problem of characterizing the asymptotic and non-asymptotic independence of functionals of a Poisson measure; see e.g. [Privault \(2012\)](#); [Rosiński and Samorodnitsky \(1999\)](#).

The remainder of the paper is organized as follows. In Section 2 we present our results in full generality. Some background material is collected in Section 3, whereas the final Section 4 contains detailed proofs, as well as some ancillary technical results.

2. Presentation of the results

We will now present an overview of the main findings of the paper. To enhance the readability of our text, we have gathered together in Section 3 definitions, notation and relevant results from the literature.

2.1. General limit theorems. Every random object considered below is defined on a suitable probability space $(\Omega, \mathcal{F}, \mathbb{P})$. The approximation results obtained in the present paper deal with (real-valued) functionals of a Poisson measure η on some Polish space $(\mathcal{Z}, \mathcal{Z})$ having non-atomic and σ -finite control μ ; see Section 3-(I). We will assume that these functionals are square-integrable random variables. To measure the distance between the distribution of a functional F of η and that of a centred Gamma random variable $G(\nu)$, we shall use the (pseudo-) metric d_3 , which is defined as follows: for every pair of square-integrable random variables X, Y , we put

$$d_3(X, Y) = \sup_{h \in \mathcal{H}^3} |\mathbb{E}[h(X)] - \mathbb{E}[h(Y)]|,$$

where $\mathcal{H}^3 := \{h \in \mathcal{C}^3 : \|h^{(j)}\|_\infty \leq 1, j \in \{1, 2, 3\}\}$ (with $h^{(j)}$ the derivative of order j of h), and where \mathcal{C}^3 is the space of thrice differentiable functions on \mathbb{R} having bounded derivatives. We notice that the topology induced by d_3 is stronger than the topology induced by convergence in distribution, which implies that if $d_3(F_n, G(\nu)) \rightarrow 0$, as $n \rightarrow \infty$, for some sequence of functionals F_n , then the distribution of F_n converges to $\bar{\Gamma}_\nu$. By a slight abuse of notation, and to stress the role of the underlying Gamma distribution, we shall often write $d_3(F, \bar{\Gamma}_\nu)$ instead of $d_3(F, G(\nu))$.

For $q \geq 1$, we write $L^2(\mu^q)$ to indicate the Hilbert space of Borel-measurable functionals on \mathcal{Z}^q that are square-integrable with respect to μ^q . We also use the following special notation: $L^2(\mu^1) = L^2(\mu)$, and $L^2_{\text{sym}}(\mu^q)$ is the subspace of $L^2(\mu^q)$ composed of those functions that are μ^q -a.e. symmetric; see Section 3-(II). Moreover, in order to simplify the notation, we use the convention that $\|\cdot\|$ and $\langle \cdot, \cdot \rangle$

stand for the norm and the scalar product in some space $L^2(\mu^q)$ whose order q will always be clear from the context.

Our first result is a quantitative estimate for $d_3(F, \bar{\Gamma}_\nu)$ in terms of the Malliavin operators D and L^{-1} , that is, the derivative operator and the pseudo-inverse of the Ornstein-Uhlenbeck generator. We recall that the derivatives DF and $DL^{-1}F$ are random elements with values in the Hilbert space $L^2(\mu)$; see Section 3-(V).

Theorem 2.1 (General Gamma bounds). *Let F be a centred and square-integrable functional of the Poisson measure η , and assume that F is in the domain of the derivative operator D . Then,*

$$\begin{aligned} d_3(F, \bar{\Gamma}_\nu) &\leq c_1 A_1(F) + c_2 A_2 + 2c_1 A_3(F) & (2.1) \\ &:= c_1 \mathbb{E} |2(F + \nu)_+ - \langle DF, -DL^{-1}F \rangle| \\ &\quad + c_2 \int_{\mathcal{Z}} \mathbb{E} [|D_z F|^2 |D_z L^{-1}F|] \mu(dz) \\ &\quad + 2c_1 \int_{\mathcal{Z}} \mathbb{E} [(D_z \mathbf{1}_{\{F > -\nu\}})(D_z F) |D_z L^{-1}F|] \mu(dz), \end{aligned}$$

with constants c_1 and c_2 given by

$$c_1 = \max(1, 1/\nu + 2/\nu^2) \quad \text{and} \quad c_2 = \max(2/3, 2/(3\nu) - 3/\nu^2 + 4/\nu^3).$$

If in addition $\mathbb{E}[\langle DF, -DL^{-1}F \rangle | F] \geq 0$ (a.s.- \mathbb{P}), then

$$A_1(F) \leq A'_1(F) := \sqrt{\mathbb{E}[(2(F + \nu) - \langle DF, -DL^{-1}F \rangle)^2]},$$

and consequently

$$d_3(F, \bar{\Gamma}_\nu) \leq c_1 A'_1(F) + c_2 A_2 + 2c_1 A_3(F) \tag{2.2}$$

Remark 2.2. (i) In (2.1), we implicitly used a ‘trajectorial’ definition of the random function $z \mapsto D_z \mathbf{1}_{\{F > -\nu\}}$, that is, we put $D_z \mathbf{1}_{\{F > -\nu\}} = \mathbf{1}_{\{F + D_z F > -\nu\}} - \mathbf{1}_{\{F > -\nu\}}$, without necessarily assuming that

$$\mathbb{E} \int_{\mathcal{Z}} (D_z \mathbf{1}_{\{F > -\nu\}})^2 \mu(dz) < \infty$$

(note that this last relation is equivalent to the fact that $\mathbf{1}_{\{F > -\nu\}}$ belongs to the set $\text{dom } D$, as defined in Section 3-(V); see Lemma 3.1). It is easily checked that

$$(D_z \mathbf{1}_{\{F > -\nu\}})(D_z F) = (\mathbf{1}_{\{F \leq -\nu < F + D_z F\}} + \mathbf{1}_{\{F + D_z F \leq -\nu < F\}}) |D_z F|,$$

in such a way that $A_3(F) \geq 0$. An effective bound on $A_3(F)$, in the case where μ is a finite measure and F is a multiple Wiener-Itô integral, is presented in Proposition 2.3.

(ii) As first done in Peccati et al. (2010), we shall often control the quantity $A_2(F)$ appearing in (2.1) by using the relation

$$\begin{aligned} A_2(F) &\leq A_4(F) \times A_5(F) \\ &:= \left(\int_{\mathcal{Z}} \mathbb{E} [|D_z F|^4] \mu(dz) \right)^{1/2} \times \left(\int_{\mathcal{Z}} \mathbb{E} [|D_z L^{-1}F|^2] \mu(dz) \right)^{1/2}. \end{aligned} \tag{2.3}$$

We also note that, if $\{F_n : n \geq 1\}$ is a sequence of random variables with bounded variances living in a fixed sum of Wiener chaoses, then the numerical sequence $n \mapsto A_5(F_n)$ is necessarily bounded.

- (iii) Theorem 2.1 should be compared with the following bound from [Nourdin and Peccati \(2009b\)](#), Theorem 3.11). Let F be a centered functional of a Gaussian measure on \mathcal{Z} with control μ , and assume that F is in the domain of the Malliavin derivative D (see [Nourdin and Peccati \(2012\)](#), Chapter 2) for relevant definitions), then there exists a constant K such that, for some adequate distance d ,

$$d(F, \bar{\Gamma}_\nu) \leq K \times \mathbb{E} |2(F + \nu)_+ - \langle DF, -DL^{-1}F \rangle|.$$

The presence of the additional term

$$\begin{aligned} & c_2 \int_{\mathcal{Z}} \mathbb{E}[|D_z F|^2 |D_z L^{-1}F|] \mu(dz) \\ & + 2c_1 \int_{\mathcal{Z}} \mathbb{E}[(D_z \mathbf{1}_{\{F > -\nu\}})(D_z F) |D_z L^{-1}F|] \mu(dz) \end{aligned}$$

in (2.1) or (2.2) is due to the characterization of the Malliavin derivative on the Poisson space as a difference operator as well as to the non-differentiability at $-\nu$ of the solution of the Stein-equation characterizing $\bar{\Gamma}_\nu$; see Section 3-(V). As proved in [Nourdin and Peccati \(2009b\)](#), Proposition 3.9, on the Gaussian-Wiener space the condition $\mathbb{E}[\langle DF, -DL^{-1}F \rangle |F] \geq 0$ (a.s.- \mathbb{P}) is satisfied for every F in the domain of D .

- (iv) Other relevant one-dimensional bounds for probabilistic approximations involving Malliavin operators on the Poisson space are proved in [Peccati et al. \(2010\)](#), dealing with normal approximations, [Peccati \(2011\)](#), dealing with the Poisson approximation of integer-valued random variables and [Eden and Viquez \(2012\)](#), focusing on absolutely continuous distributions whose support is given by the real line. See [Bourguin and Peccati \(2012\)](#); [Peccati and Zheng \(2010\)](#) for several multidimensional extensions.

As announced, we conclude the present section with a useful bound on the quantity $A_3(F)$, in the case where $F = I_q(f)$ equals a multiple Wiener-Itô integral and the control measure μ is finite. At the cost of a heavier notation, our techniques could suitably be modified in order to deal with the case of a random variable F having a finite chaotic expansion.

Proposition 2.3. *Let the control measure μ be finite, and consider $F = I_q(f)$, where $q \geq 2$ and $f \in L^2_{\text{sym}}(\mu^q)$. We assume that (i) $\mathbb{E} \int_{\mathcal{Z}} (D_z F)^4 \mu(dz) < \infty$, that (ii) the random function*

$$\mathcal{Z} \ni z \mapsto D_z F |D_z F| := v(z)$$

is such that $v(z) \in \text{dom } D$ for $\mu(dz)$ -almost every z , and satisfies

$$\mathbb{E} \int_{\mathcal{Z}} \int_{\mathcal{Z}} (D_{z_2} v(z_1))^2 \mu(dz_1) \mu(dz_2) < \infty.$$

Then, defining $A_3(F)$ as in (2.1), one has the bound

$$\begin{aligned} \frac{q}{2\sqrt{2}}A_3(F) &\leq \sqrt{\mathbb{E} \int_{\mathcal{Z}} (D_z F)^4 \mu(dz)} \\ &\quad + \sqrt{\mathbb{E} \int_{\mathcal{Z}} \int_{\mathcal{Z}} (D_{z_2} D_{z_1} F)^2 (D_{z_1} F)^2 \mu(dz_1) \mu(dz_2)} \\ &\quad + \sqrt{\mathbb{E} \int_{\mathcal{Z}} \int_{\mathcal{Z}} (D_{z_2} D_{z_1} F)^4 \mu(dz_1) \mu(dz_2)}. \end{aligned} \quad (2.4)$$

Remark 2.4. (i) Another way of controlling the term $A_3(F)$, whenever F has a finite chaotic expansion, is discussed in Schulte (2012). One should note that, albeit our proof of Proposition 2.3 also starts with an integration by parts formula, our strategy for controlling the term $A_3(F)$ is significantly different. Indeed, our approach is based on isometric formulae for divergence operators, whereas Schulte (2012) uses a direct estimation consisting in controlling $|DF|$ by a random function having a finite chaotic expansion. When applied to our framework in the case $q > 2$, the technique used in Schulte (2012) leads to expressions involving contractions of the absolute value of the kernel f , therefore producing bounds that are systematically larger than ours. When applied to the case $q = 2$, the strategy adopted in Schulte (2012) leads to slower rates of convergence, but allows in principle to dispense with the assumption that the underlying control measure has finite mass. Since all our applications concern sequences of control measures having a finite mass, and for the sake of conciseness, we will omit a formal discussion of this fact.

- (ii) From the standpoint of geometric applications, focusing on Poisson measures having a finite control is barely a restriction. Indeed, the kind of geometric limit theorems we are interested in typically involve either functionals of a Poisson measure having a finite control, whose total mass asymptotically explodes (like the ones we consider in the applications developed later in the paper), or functionals of the restriction of a Poisson measure to a finite window with growing volume; see e.g. Bourguin and Peccati (2012); Decreusefond et al. (2011); Lachièze-Rey and Peccati (2013a,b); Last et al. (2012); Peccati (2011); Reitzner and Schulte (2012+); Schulte and Thäle (2012) for a recent collection of distinguished examples.

2.2. Simplified estimates for supports contained in a half-line. The applications we are interested in require that we consider random variables possibly taking values in the half-line $(-\infty, -\nu)$, in such a way that the rather unusual term $A_3(F)$ cannot be dispensed with. However, if one is only interested in measuring the distance between $\bar{\Gamma}_\nu$ and the law of a random variable with support in $[-\nu, +\infty)$, then the statement of Theorem 2.1 can be significantly simplified, since in this case the term $A_3(F)$ disappears. In particular, whenever the law of F satisfies these requirements, the finiteness of the measure μ does not play any role. This point is made clear in the next statement whose easy proof is left to the reader.

Proposition 2.5. *Let F be a centered square-integrable functional of the random measure η . Assume that the law of F has support in $[-\nu, +\infty)$ and that F is in the*

domain of the derivative operator D . Then, the bound (2.1) holds with $A_3(F) = 0$. If moreover $\mathbb{E}[\langle DF, -DL^{-1}F \rangle | F] \geq 0$ (a.s.- \mathbb{P}), then the estimate (2.2) holds with $A_3(F) = 0$.

2.3. General results for sequences of multiple integrals. We now focus on the following setup. Let $(\mathcal{Z}, \mathcal{Z})$ be a fixed Polish space as above, and $\{\eta_n : n \geq 1\}$ be a sequence of Poisson random measures on $(\mathcal{Z}, \mathcal{Z})$, such that, for each n , the non-atomic control measure μ_n of η_n is finite. In view of applications, we allow that $\mu_n(\mathcal{Z}) \rightarrow \infty$, as $n \rightarrow \infty$. For a given even integer $q \geq 2$, we consider a sequence $\{I_q(f_n) : n \geq 1\}$ of multiple Wiener-Itô stochastic integrals with the following characteristics: (a) $\{f_n : n \geq 1\} \subset L^2_{\text{sym}}(\mu_n^q)$ is composed of kernels satisfying the technical assumptions stated in Section 3-(VIII) below, and (b) for every $n \geq 1$, the integral $I_q(f_n)$ is realized with respect to the compensated Poisson measure $\hat{\eta}_n = \eta_n - \mu_n$. The next theorem characterizes the convergence of the distribution of $I_q(f_n)$, as $n \rightarrow \infty$, to the limit law $\bar{\Gamma}_\nu$. The set of analytic conditions appearing below is expressed in terms of (possibly symmetrized) contraction kernels, whose definition is provided in Section 3-(VI). Observe in particular that $f_n \star_q^0 f_n = f_n^2$.

Theorem 2.6 (Gamma limits in the Poisson-Wiener chaos). *Let the above assumptions and notation prevail (in particular, μ_n is a finite measure for every n), let $q \geq 2$ be an even integer and let $\{f_n : n \geq 1\} \subset L^2_{\text{sym}}(\mu_n^q)$ be such that $\lim_{n \rightarrow \infty} q! \|f_n\|^2 = 2\nu$, and suppose that the technical conditions of Section 3-(VIII) are satisfied. Assume in addition that*

$$\begin{aligned} \lim_{n \rightarrow \infty} \|f_n \star_r^\ell f_n\| &= 0 \\ \text{and } \lim_{n \rightarrow \infty} \|f_n \tilde{\star}_{q/2}^{q/2} f_n - c_q f_n\| &= 0 \quad \text{with } c_q = \frac{4}{\left(\frac{q}{2}\right)! \left(\frac{q}{2}\right)^2} \end{aligned} \quad (2.5)$$

for all pairs (r, ℓ) such that either $r = q$ and $\ell = 0$, or $r \in \{1, \dots, q\}$, $\ell \in \{1, \dots, \min(r, q-1)\}$ and r and ℓ are not equal to $q/2$ at the same time. Then, the distribution of $I_q(f_n)$ converges to $\bar{\Gamma}_\nu$ as $n \rightarrow \infty$. Moreover, for some positive finite constant K independent of n ,

$$\begin{aligned} d_3(I_q(f_n), \bar{\Gamma}_\nu) &\leq c_1 A_1(I_q(f_n)) + c_2 A_4(I_q(f_n)) \times A_5(I_q(f_n)) + 2c_1 A_3(I_q(f_n)) \\ &\leq K \times \max \left\{ |q! \|f_n\|^2 - 2\nu|; \|f_n \star_p^p f_n\|; \right. \\ &\quad \left. \|f_n \star_r^\ell f_n\|^{1/2}; \|f_n \tilde{\star}_{q/2}^{q/2} f_n - c_q f_n\| \right\} \rightarrow 0, \end{aligned} \quad (2.6)$$

where we have used the notation introduced in (2.1)-(2.3), and the maximum is taken over all $p = 1, \dots, q-1$ such that $p \neq q/2$ and all (r, ℓ) such that $r \neq \ell$ and either $r = q$ and $\ell = 0$, or $r \in \{1, \dots, q\}$ and $\ell \in \{1, \dots, \min(r, q-1)\}$.

Example 2.7. (i) Assume $q = 2$. Then, $c_2 = 1$ and the maximum in (2.6) is taken over the following four quantities:

$$|2\|f_n\|^2 - 2\nu|, \quad \|f_n \star_2^0 f_n\|^{1/2}, \quad \|f_n \star_2^1 f_n\|^{1/2}, \quad \|f_n \tilde{\star}_1^1 f_n - f_n\|.$$

(ii) Assume $q = 4$. Then, $c_4 = 1/18$ and the maximum in (2.6) is taken over the following ten quantities:

$$\begin{aligned} &|2\|f_n\|^2 - 2\nu|, \quad \|f_n \star_1^1 f_n\|, \quad \|f_n \star_4^0 f_n\|^{1/2}, \quad \|f_n \star_4^1 f_n\|^{1/2}, \quad \|f_n \star_4^2 f_n\|^{1/2}, \\ &\|f_n \star_4^3 f_n\|^{1/2}, \quad \|f_n \star_3^1 f_n\|^{1/2}, \quad \|f_n \star_3^2 f_n\|^{1/2}, \quad \|f_n \star_2^1 f_n\|^{1/2}, \end{aligned}$$

and $\|f_n \star_1^{-1} f_n - 18^{-1} f_n\|$, where we have used the fact that $\|f_n \star_1^1 f_n\| = \|f_n \star_3^3 f_n\|$.

Remark 2.8. (i) Under the assumptions in the statement, one has that the sequence

$$A_5(I_q(f_n)) := \left(\int_{\mathcal{Z}} \mathbb{E}[|D_z L^{-1} I_q(f_n)|^2] \mu_n(dz) \right)^{1/2}$$

is such that

$$A_5(I_q(f_n))^2 = (q - 1)! \|f_n\|^2 \rightarrow \frac{2\nu}{q} > 0 \quad \text{as } n \rightarrow \infty.$$

It follows that our inequality (2.6) not only provides an analytic bound in the distance d_3 , but also ensures that the three numerical sequences $\{A_1(I_q(f_n)) : n \geq 1\}$, $\{A_3(I_q(f_n)) : n \geq 1\}$ and $\{A_4(I_q(f_n)) : n \geq 1\}$ (all related to Malliavin operators) converge to zero. This fact is crucial when dealing with the multidimensional results discussed in Section 2.6. An analogous remark applies to Proposition 2.9 and Theorem 2.13 below.

- (ii) Similar conditions (only involving contractions of the type \star_r^r , with $r = 1, \dots, q - 1$) in the case of multiple integrals with respect to a Gaussian measure can be found in Nourdin and Peccati (2009a, Theorem 1.2). Non-central results of a similar flavor, in the context of free probability and multiple integrals with respect to a free Brownian motion, are proved in Nourdin and Peccati (2013+).
- (iii) We were able to deduce meaningful conditions for Gamma approximations only in the case of an *even* integer $q \geq 2$. However, unlike in the Gaussian case (see Nourdin and Peccati (2009a, Remark 1.3)), in a Poisson framework one cannot exclude a priori the existence of a sequence of multiple integrals of odd order converging to a limiting Gamma distribution. We prefer to consider this issue as a separate problem, and keep it as an open direction for future research.
- (iv) In the estimate (2.6), and in contrast to the main bounds on normal approximations proved in Peccati et al. (2010), norms of the type $\|f_n \star_r^\ell f_n\|$, $r \neq \ell$, appear under a square root. This phenomenon seems unavoidable, and it is directly related to the presence of cross terms arising from the specific form of the Stein equation associated with the Gamma distribution.

The following statement shows that condition (2.5) might take a particularly attractive form in the case of double Poisson integrals. This will be used in order to prove the results presented in Section 2.4, dealing with the Gamma approximation of degenerate U -statistics.

Proposition 2.9 (Three moments suffice for Gamma approximations).

Let the control measures $\{\mu_n : n \geq 1\}$ be finite, let $q = 2$ and let $\{f_n : n \geq 1\} \subset L^2_{\text{sym}}(\mu_n^2)$ be such that $\lim_{n \rightarrow \infty} \mathbb{E}[I_2^2(f_n)] = \lim_{n \rightarrow \infty} 2\|f_n\|^2 = 2\nu$, and such that the technical conditions of Section 3-(VIII) are satisfied. Assume in addition that $\int_{\mathcal{Z}} f_n^4 d\mu_n^2 \rightarrow 0$ and that $\mathbb{E}[I_2^4(f_n)] < \infty$ for every n . Then, condition (2.5) is verified if and only if

$$\mathbb{E}[I_2^4(f_n)] - 12\mathbb{E}[I_2^3(f_n)] \longrightarrow 12\nu^2 - 48\nu \quad \text{as } n \rightarrow \infty. \tag{2.7}$$

In particular, if the sequence F_n^4 is uniformly integrable, then (2.5) and (2.7) are both necessary and sufficient in order to have that the distribution of F_n converges to $\bar{\Gamma}_\nu$ in the sense of the distance d_3 .

2.4. *An extension of de Jong's theorem for degenerate U -statistics.* In the present and the subsequent section, we shall work within the following framework. We fix an integer $d \geq 1$, and let $\mathbf{Y} = \{Y_i : i \geq 1\}$ be a sequence composed of i.i.d. random variables with values in \mathbb{R}^d , whose common distribution has a density $p(x)$ with respect to the Lebesgue measure on \mathbb{R}^d (written dx). The sequence $\{N(n) : n \geq 1\}$ of integer-valued random variables is independent of \mathbf{Y} and such that, for every n , $N(n)$ has a Poisson distribution with parameter n . It is well-known that, in this setting, the random point measure

$$\eta_n := \sum_{i=1}^{N(n)} \delta_{Y_i} \quad (2.8)$$

(where δ_y represents the Dirac mass at y) is a Poisson measure on $\mathcal{Z} = \mathbb{R}^d$ (equipped with the standard Borel σ -field $\mathcal{B}(\mathbb{R}^{\otimes d})$) with control measure $\mu_n(dx) = np(x)dx$. We shall also use the shorthand notation $\mu(dx) := \mu_1(dx) = p(x)dx$.

Our aim below is to provide a Gamma-type counterpart to a famous theorem by P. de Jong, proved in de Jong (1987), involving sequences of degenerate U -statistics of order 2. We stress that the results contained in de Jong (1987) have later been extended to degenerate U -statistics of a general order; see Bhattacharya and Ghosh (1992); de Jong (1990). Albeit our method clearly applies to these general objects, we prefer here to focus on U -statistics of order 2, in order to obtain neater statements and to emphasize the method over technical details. We start with some useful definitions.

Definition 2.10 (U -statistics). (i) Let $k \geq 2$, and let $h : \mathbb{R}^q \rightarrow \mathbb{R}$ be a symmetric kernel such that $h \in L_{\text{sym}}^1(\mu^k)$. The (symmetric) U -statistic of order k based on h and on the sample $\{Y_1, \dots, Y_m\}$ (where $m \geq k$ is some integer) is the random variable

$$U_m(h, \mathbf{Y}) = \sum_{i_1, \dots, i_k=1}^m \neq h(Y_{i_1}, \dots, Y_{i_k}), \quad (2.9)$$

where the symbol $\sum \neq$ indicates that the sum is taken over all vectors (i_1, \dots, i_k) such that $i_j \neq i_\ell$ for every $j \neq \ell$.

(ii) Fix $k \geq 2$ and let $U_m(h, \mathbf{Y})$ be a symmetric U -statistic as in (2.9). The *Hoeffding rank* of $U_m(h, \mathbf{Y})$ is the smallest integer $1 \leq q \leq k$ such that $\mathbb{E}[h(Y_1, \dots, Y_k) | Y_1, \dots, Y_{q-1}] = 0$ (a.s.- \mathbb{P}) and $\mathbb{E}[h(Y_1, \dots, Y_k) | Y_1, \dots, Y_q] \neq 0$, where $\mathbb{E}[h(Y_1, \dots, Y_k) | Y_1, \dots, Y_0] := \mathbb{E}[h(Y_1, \dots, Y_k)]$. A U -statistic of order k with Hoeffding rank equal to k is said to be *completely degenerate*. In other words, a U -statistic such as (2.9) is completely degenerate if h is a non-zero kernel verifying

$$\int_{\mathbb{R}} h(x, y_1, \dots, y_{k-1}) p(x) dx = 0 \quad (\mu^{k-1} - \text{a.e.}).$$

(iii) A collection of random variables $\{F_n : n \geq 1\}$ is said to be a sequence of *geometric U -statistics* of order k , if there exists a kernel $h \in L_{\text{sym}}^1(\mu^k)$ such

that

$$F_n = U_{N(n)}(h, \mathbf{Y}), \quad n \geq 1,$$

where $\{N(n) : n \geq 1\}$ is the independent Poisson sequence introduced above.

Before presenting the main result of this section, and in order to make the connection with our general framework more transparent, we shall recall an important finding from [Reitzner and Schulte \(2012+, Lemma 3.5 and Theorem 3.6\)](#), stating that Poissonized U -statistics of order k live inside the sum of the first $k + 1$ Wiener chaoses associated with the Poisson measure η_n . The proof heavily relies on results by [Last and Penrose \(2011\)](#).

Lemma 2.11 (Reitzner and Schulte). *Consider a kernel $h \in L^1_{\text{sym}}(\mu^k)$ such that the corresponding Poissonized U -statistic $U_{N(n)}(h, \mathbf{Y})$ is square-integrable. Then, h is necessarily in $L^2_{\text{sym}}(\mu^k)$, and $U_{N(n)}(h, \mathbf{Y})$ admits a chaotic representation of the type*

$$U_{N(n)}(h, \mathbf{Y}) = \mathbb{E}[U_{N(n)}(h, \mathbf{Y})] + \sum_{i=1}^k n^{k-i} I_i(h_i)$$

where I_i indicates a multiple Wiener-Itô integral of order i with respect to the compensated Poisson measure $\hat{\eta}_n = \eta_n - \mu_n$, defined according to [\(2.8\)](#), and

$$h_i(z_1, \dots, z_i) = \binom{k}{i} \int_{\mathcal{Z}^{k-i}} h(z_1, \dots, z_i, \bullet) \mu^{k-i}(d\bullet), \quad (z_1, \dots, z_i) \in \mathcal{Z}^i, \quad (2.10)$$

where the bullet “ \bullet ” stands for a packet of $k - i$ variables that are integrated with respect to μ^{k-i} . In particular, $h = h_k$ and the projection h_i is in $L^2_{\text{sym}}(\mu^i)$ for each $1 \leq i \leq k$.

The following statement corresponds to the main result proved by [de Jong \(1987\)](#), in the special case of symmetric U -statistics of order 2 (note that the assumption that the underlying kernels have finite moments of order four is only implicit in de Jong’s work). Given positive sequences $a_n, b_n, n \geq 1$, we write $a_n \approx b_n$ whenever $\lim_{n \rightarrow \infty} a_n/b_n = 1$.

Theorem 2.12 (de Jong). *Let $\{h_n : n \geq 1\}$ be a sequence of non-zero elements of $L^4_{\text{sym}}(\mu^2)$. Define $F_n = U_n(h_n, \mathbf{Y})$ and assume that F_n is completely degenerate. Then, one has that $\sigma^2(n) := \text{Var}(F_n) \approx 2n^2 \mathbb{E}[h_n(Y_1, Y_2)^2]$, and the fourth moment condition*

$$\lim_{n \rightarrow \infty} \frac{\mathbb{E}[F_n^4]}{\sigma(n)^4} = 0,$$

implies that, as $n \rightarrow \infty$, the sequence $\tilde{F}_n := F_n/\sigma(n)$ converges in distribution to a standard Gaussian random variable.

The following statement consists of two parts. Part (A) is a quantitative extension of [Theorem 2.12](#) based on a direct study of the fourth moments of the Poissonized U -statistic, whereas part (B) is a Gamma-type extension of de Jong’s theorem which is directly based on the results discussed in [Section 2.1](#). Apart from [de Jong \(1987\)](#), our findings should be compared with the seminal work by [Jammalamadaka and Janson \(1986\)](#), about the normal and Poisson approximation of U -statistics of order two. To our knowledge, the forthcoming [Theorem 2.13](#) is the

first quantitative extensions of the de Jong theorem, also dealing with the non-normal approximation of general degenerate U -statistics. Moreover, we would like to emphasize that our proof of Part (A) is shorter and more transparent than the one presented in the original work [de Jong \(1987\)](#) (one should note that, however, our methods only allow us to deal with symmetric U -statistics). Recall that the *Wasserstein distance* between the laws of two integrable random variables X, Y is given by

$$d_W(X, Y) := \sup_{h \in \text{Lip}(1)} |\mathbb{E}[h(X)] - \mathbb{E}[h(Y)]|,$$

where $\text{Lip}(1)$ is the set of Lipschitz functions $h : \mathbb{R} \rightarrow \mathbb{R}$ with a Lipschitz constant ≤ 1 . Recall that, in the framework of this section, $\mathcal{Z} = \mathbb{R}^d$.

Theorem 2.13 (Extended de Jong theorem). *Let $\{h_n : n \geq 1\}$ be a sequence of non-zero elements of $L_{\text{sym}}^4(\mu^2)$ such that*

$$\sup_n \frac{\int_{\mathcal{Z}} h_n^4 d\mu_n^2}{\left(\int_{\mathcal{Z}} h_n^2 d\mu_n^2\right)^2} < \infty.$$

Put $F_n = U_n(h_n, \mathbf{Y})$ and $F'_n = U_{N(n)}(h_n, \mathbf{Y})$, and assume that these U -statistics are completely degenerate. Then, $\sigma(n)^2 := \text{Var}(F_n) \approx \text{Var}(F'_n) = 2n^2 \mathbb{E}[h_n(Y_1, Y_2)^2]$, and the following two points (A) and (B) hold.

(A) *If*

$$\frac{\mathbb{E}[(F'_n)^4]}{\sigma(n)^4} \rightarrow 0 \quad \text{as } n \rightarrow \infty, \quad (2.11)$$

then both $\tilde{F}_n := F_n/\sigma(n)$ and $\tilde{F}'_n := F'_n/\sigma(n)$ converge in distribution to a standard Gaussian random variable N . Moreover, there exists a universal finite constant K , independent of n , such that, as $n \rightarrow \infty$,

$$d_W(\tilde{F}'_n, N) \leq K \times B_n \rightarrow 0, \quad (2.12)$$

$$d_W(\tilde{F}_n, N) \leq K \times (B_n + n^{-1/4}) \rightarrow 0, \quad (2.13)$$

with $B_n := \sigma(n)^{-2} \max \left\{ \left(\int_{\mathcal{Z}} h_n^4 d\mu_n^2 \right)^{1/2}; \|h_n \star_1 h_n\|; \|h_n \star_2 h_n\| \right\}$.

(B) *If $\int_{\mathcal{Z}} h_n^4 d\mu_n^2 \rightarrow 0$ and there exists $\nu > 0$ such that $\sigma(n)^2 \rightarrow 2\nu$, and*

$$\mathbb{E}[(F'_n)^4] - 12\mathbb{E}[(F'_n)^3] \rightarrow 12\nu^2 - 48\nu \quad \text{as } n \rightarrow \infty, \quad (2.14)$$

then both F_n and F'_n converge in distribution to a random variable $G(\nu)$, which has distribution $\bar{\Gamma}_\nu$. Moreover, there exists a universal constant $K > 0$ such that, as $n \rightarrow \infty$,

$$d_3(F'_n, \bar{\Gamma}_\nu) \leq c_1 A_1(F'_n) + c_2 A_4(F'_n) \times A_5(F'_n) \leq K \times C_n \rightarrow 0, \quad (2.15)$$

$$d_3(F_n, \bar{\Gamma}_\nu) \leq K \times (C_n + n^{-1/4}) \rightarrow 0, \quad (2.16)$$

with

$$C_n := \max \left\{ |2\|h_n\|^2 - 2\nu|; \left(\int_{\mathcal{Z}} h_n^4 d\mu_n^2 \right)^{1/4}; \|h_n \star_2 h_n\|^{1/2}; \|h_n \tilde{\star}_1 h_n - h_n\| \right\},$$

and we have used the notation introduced in (2.1)–(2.3).

Remark 2.14. Our proof of Theorem 2.13 shows indeed that the quantity B_n (reps. C_n) in the statement converges to zero if and only if the asymptotic condition (2.11) (resp. (2.14)) is verified.

2.5. *Gamma convergence of geometric U-statistics: characterization and bounds.* As anticipated, the aim of this section is to apply the main estimates of the present paper in order to characterize the class of geometric U-statistics based on \mathbf{Y} converging in distribution towards a Gamma random variable. Since our analysis is based on Theorem 2.13, our results will provide explicit estimates on the speed of convergence. We refer the reader to Dynkin and Mandelbaum (1983); Rubin and Vitale (1980) for some classic references on the subject and to Lachièze-Rey and Peccati (2013b); Reitzner and Schulte (2012+) for a discussion of several recent developments. We let the notation and assumptions of the previous section prevail and recall that a Gaussian measure G on $(\mathbb{R}^d, \mathcal{B}(\mathbb{R})^{\otimes d})$, with control $\mu(dx) = p(x)dx$, is a centred Gaussian family of the type

$$G = \{G(B) : B \in \mathcal{B}(\mathbb{R})^{\otimes d}, \mu(B) < \infty\}$$

such that, for every $m \geq 1$ and every $B_1, \dots, B_m \in \mathcal{B}(\mathbb{R})^{\otimes d}$ with $\mu(B_i) < \infty$ ($i = 1, \dots, m$), the vector $(G(B_1), \dots, G(B_m))$ has an m -dimensional joint Gaussian distribution with covariance matrix $\mathbb{E}[G(B_i)G(B_j)] = \mu(B_i \cap B_j)$.

The next statement combines findings from Lachièze-Rey and Peccati (2013b, Section 7) (point (i)) with a classic characterization of elements in the second Wiener chaos of a Gaussian measure (point (ii); see Nourdin and Peccati (2012, Section 2.7.4) for more details.

Proposition 2.15. *Let $k \geq 2$ and let $h \in L^1_{\text{sym}}(\mu^k)$ be a non-zero kernel such that the U-statistic $F'_n := U_{N(n)}(h, \mathbf{Y})$ is square-integrable for every n , and has Hoeffding rank equal to 2. For $n \geq 1$, define also the standardized U-statistic $\tilde{F}'_n = n^{1-k}F'_n$.*

- (i) *For every n , there exists a sequence of double integrals $I_2(f_n)$ (each realized with respect to the compensated Poisson measure $\eta_n - \mu_n$) such that, as $n \rightarrow \infty$, $\mathbb{E}[(\tilde{F}'_n - I_2(f_n))^2] \rightarrow 0$. Moreover, \tilde{F}'_n converge in distribution to $I_2^G(h_2)$, where I_2^G indicates a double Wiener-Itô integral with respect to the Gaussian measure G , and h_2 is defined according to (2.10). The same convergence takes place for the de-Poissonized U-statistics $\tilde{F}_n = n^{1-k}F_n$, where $F_n := U_n(h, \mathbf{Y})$.*
- (ii) *The random variable $I_2^G(h_2)$ cannot be Gaussian. Moreover, assume that $I_2^G(h_2)$ follows a $\bar{\Gamma}_\nu$ -distribution. Then, necessarily, $\nu \in \{1, 2, \dots\}$ and there exists an orthonormal system $\{e_1, \dots, e_\nu\} \subset L^2(\mu)$ such that $h_2 = \sum_{i=1}^\nu e_i \otimes e_i$ and $\mathbb{E}[e_i(Y_1)] = 0$ for $i = 1, \dots, \nu$.*

Remark 2.16. Let $k = 2$, and consider $h_2 = \sum_{i=1}^\nu e_i \otimes e_i$, as in the statement of Proposition 2.15-(ii). Then, it is easily seen (by a direct computation) that

$$U_n(h_2, \mathbf{Y}) = \sum_{i=1}^\nu \left[\left(\sum_{k=1}^n e_i(Y_k) \right)^2 - \sum_{k=1}^n e_i(Y_k)^2 \right].$$

The fact that the distributions of \tilde{F}'_n and \tilde{F}_n converge to $\bar{\Gamma}_\nu$ is therefore a direct consequence of the usual multidimensional central limit theorem and of the law of large numbers.

The next statement is a quantitative counterpart to Proposition 2.15-(ii), containing in particular estimates involving Malliavin operators. Such estimates will be put into use in the Examples 2.22–2.26 below, where the asymptotic behavior of a U -statistic such as $U_{N(n)}(h_2, \mathbf{Y})$ is studied within the framework of hybrid convergence in random graphs, random flat and random simplex models.

Theorem 2.17 (Bounds on Gamma convergence). *Let the assumptions and notation of Proposition 2.15 prevail. Assume moreover that $I_2^G(h_2)$ has distribution $\bar{\Gamma}_\nu$ for some $\nu = 1, 2, \dots$, and also that $\{e_1, \dots, e_\nu\} \subset L^4(\mu)$, where the orthonormal system $\{e_1, \dots, e_\nu\}$ is defined in Proposition 2.15-(ii). Then, there exists a finite constant K , independent of n , such that*

$$\begin{aligned} d_3(\tilde{F}'_n, \bar{\Gamma}_\nu) &\leq c_1 A_1(\tilde{F}'_n) + c_2 A_4(\tilde{F}'_n) \times A_5(\tilde{F}'_n) \leq K \times n^{-1/4}, \\ d_3(\tilde{F}_n, \bar{\Gamma}_\nu) &\leq K \times n^{-1/4}, \end{aligned}$$

where in the first inequality we used the notation defined in (2.1)–(2.3)

Example 2.18. (i) Let g_1, g_2 be two orthonormal elements of $L^2(\mu)$ such that $g_1, g_2 \in L^4(\mu)$ and $\int_{\mathcal{Z}} g_1(z) \mu(dz) = \int_{\mathcal{Z}} g_2(z) \mu(dz)$. We stress that we do not require that g_1, g_2 have disjoint supports. Then, the kernel

$$\begin{aligned} h_2(z_1, z_2) &= \frac{1}{2}(g_1 - g_2) \otimes (g_1 - g_2)(z_1, z_2) \\ &= \frac{g_1(z_1) - g_2(z_1)}{\sqrt{2}} \times \frac{g_1(z_2) - g_2(z_2)}{\sqrt{2}} \end{aligned}$$

is such that the corresponding U -statistics of order two $F_n := U_n(h_2, \mathbf{Y})$ and $F'_n := U_{N(n)}(h_2, \mathbf{Y})$ are completely degenerate, and both converge in distribution to $\bar{\Gamma}_1$, with an upper bound of order $n^{-1/4}$ on the rate of convergence.

- (ii) As an example of a pair (g_1, g_2) verifying the requirements at Point (i), one can take $g_1 = \sqrt{2} \mathbf{1}_A$ and $g_2 = \sqrt{2} \mathbf{1}_B$, where $\{A, B\}$ is a measurable partition of \mathcal{Z} such that $\mu(A) = \mu(B) = 1/2$. Considering the case $d = 1$, $p(\cdot) = \frac{1}{2} \mathbf{1}_{(-1,1)}(\cdot)$, $g_1(z) = \sqrt{2} \mathbf{1}_{(0,1)}(z)$ and $g_2(z) = \sqrt{2} \mathbf{1}_{(-1,0)}(z)$, one obtains a kernel h_2 with support in $(-1, 1)^2 \setminus \{(0, 0)\}$ and such that $h_2(z_1, z_2) = 1$ if $z_1 z_2 > 0$, and $h_2(z_1, z_2) = -1$ if $z_1 z_2 < 0$. In this way, one recovers the non-central result discussed by Reitzner and Schulte in Reitzner and Schulte (2012+, end of Section 5.1).

2.6. Multivariate extensions and hybrid convergence. We describe here three multivariate extensions of the results in Section 2.1. The first two results can be seen as partial analogues on the Poisson space of Nourdin and Rosinski (2012+, Theorem 4.4) – for the multivariate Gamma convergence – and Nourdin and Rosinski (2012+, Theorem 4.5) – for the hybrid convergence – both concerning sequences of multiple integrals with respect to a Gaussian measure. Observe that the method used in Nourdin and Rosinski (2012+) is based on new criteria for asymptotic independence of multiple integrals. These criteria are not available on the Poisson space.

For this reason, our approach is different and will be in the spirit of the “interpolation method” used in Bourguin and Peccati (2012). Such an interpolation method will be also used to deduce our third result, concerning hybrid Poisson/Gamma convergence. As already pointed out, the general problem of characterizing independence is rather well-understood in a Gaussian framework (cf. Kallenberg (1991); Üstünel and Zakai (1989, 1990)), while the topic is still largely open in the context Poisson measures; see Privault (2012); Rosiński and Samorodnitsky (1999).

Remark 2.19. As before, we consider the framework of a sequence of Poisson measures $\{\eta_n : n \geq 1\}$ (on some Polish space $(\mathcal{Z}, \mathcal{Z})$), each having a finite with non-atomic control measure μ_n . For the entire section, I_q denotes the multiple Wiener-Itô integral, of order q , with respect to one of the compensated measures $\hat{\eta}_n = \eta_n - \mu_n$ (the concerned index n will always coincide with the index of the integrated function, for instance: $I_q(f_n)$ indicates the multiple integral of order q of f_n with respect to $\hat{\eta}_n$).

Let $d \geq 1$ be a fixed integer, let $\nu_1, \dots, \nu_d > 0$ and let (G_1, \dots, G_d) be a vector consisting of independent random variables such that G_i has the centred Gamma distribution $\bar{\Gamma}_{\nu_i}$. Further let $2 \leq q_1 < q_2 < \dots < q_d$ be even integers satisfying $2q_i \neq q_j$ for any $i \neq j$ and let for $i \in \{1, \dots, d\}$, $\{f_n^{(i)} : n \geq 1\}$ be a sequence of kernels such that $f_n^{(i)} \in L^2_{\text{sym}}(\mu_n^{q_i})$, satisfying in addition the technical conditions of Section 3-(VIII). The next result deals with the announced multivariate Gamma convergence. We emphasize that we do not need further conditions on asymptotic covariances due to our assumption that all multiple integrals have different orders.

Theorem 2.20 (Multivariate Gamma convergence). *Let the above notation and conditions prevail. For any $i \in \{1, \dots, d\}$ assume that $\lim_{n \rightarrow \infty} q_i! \|f_n^{(i)}\|^2 = 2\nu_i$ and that*

$$\lim_{n \rightarrow \infty} \|f_n^{(i)} \star_{r_i}^{\ell_i} f_n^{(i)}\| = 0$$

$$\text{and } \lim_{n \rightarrow \infty} \|f_n^{(i)} \tilde{\star}_{q_i/2}^{q_i/2} f_n^{(i)} - c_{q_i} f_n^{(i)}\| = 0 \quad \text{with } c_{q_i} = \frac{4}{\left(\frac{q_i}{2}\right)! \left(\frac{q_i}{2}\right)^2}$$

for all pairs (r_i, ℓ_i) such that either $r_i = q_i$ and $\ell_i = 0$, or $r_i \in \{1, \dots, q_i\}$, $\ell_i \in \{1, \dots, \min(r_i, q_i - 1)\}$ and r_i and ℓ_i not equal to $q_i/2$ at the same time. Then, $(I_{q_1}(f_n^{(1)}), \dots, I_{q_d}(f_n^{(d)}))$ converges in distribution to (G_1, \dots, G_d) as $n \rightarrow \infty$.

We go one step further and turn to an extension of Theorem 2.20 where we consider convergence of a random vector of multiple integrals to a hybrid random vector whose components are independent and in part centered Gamma and in part standard Gaussian random variables. A similar setting with Poisson random variables instead

of centred Gamma ones has recently been studied in Bourguin and Peccati (2012). However, we would like to emphasize that, in contrast to Bourguin and Peccati (2012), here both distributions considered in the target vector are absolutely continuous with respect to the Lebesgue measure on the real line.

To formulate our result on the Gamma/Gaussian hybrid convergence, let $d_1, d_2 \geq 1$ be fixed integers, let $\nu_1, \dots, \nu_{d_1} > 0$ and let $(G_1, \dots, G_{d_1}, N_{d_1+1}, \dots, N_{d_1+d_2})$ be a vector consisting of independent random variables such that G_i has distribution $\bar{\Gamma}_{\nu_i}$ for $i \in \{1, \dots, d_1\}$ and N_i has a standard Gaussian distribution for $i \in \{d_1 + 1, \dots, d_1 + d_2\}$.

$1, \dots, d_1 + d_2$. Further let $2 \leq q_1, q_2, \dots, q_{d_1+d_2}$ be integers such that the following constraints are verified: (a) $q_1 < \dots < q_{d_1}$ and, in general, $q_i \neq q_j$ for every $1 \leq i \neq j \leq d_1 + d_2$, (b) q_1, \dots, q_{d_1} are even integers, and (c) there is no pair $(i, j) \in \{1, \dots, d_1 + d_2\}^2$ such that $i \in \{1, \dots, d_1\}$ and $2q_i = q_j$. For $i \in \{1, \dots, d_1 + d_2\}$, we let $\{f_n^{(i)} : n \geq 1\}$ be a sequence of symmetric and square-integrable kernels such that $f_n^{(i)} \in L_{\text{sym}}^2(\mu_n^{q_i})$, satisfying moreover the technical conditions stated in Section 3-(VIII).

Theorem 2.21 (Gamma/Normal hybrid convergence). *Let the above notation and conditions prevail. Assume that $\lim_{n \rightarrow \infty} q! \|f_n^{(i)}\|^2 = 2\nu_i$ whenever $i \in \{1, \dots, d_1\}$ and that $\lim_{n \rightarrow \infty} q_i! \|f_n^{(i)}\|^2 = 1$ whenever $i \in \{d_1 + 1, \dots, d_2\}$. Furthermore, for any $i \in \{1, \dots, d_1\}$ suppose that*

$$\lim_{n \rightarrow \infty} \|f_n^{(i)} \star_{r_i}^{\ell_i} f_n^{(i)}\| = 0$$

and $\lim_{n \rightarrow \infty} \|f_n^{(i)} \star_{q_i/2}^{\sim q_i/2} f_n^{(i)} - c_{q_i} f_n^{(i)}\| = 0$ with $c_{q_i} = \frac{4}{(\frac{q_i}{2})! (\frac{q_i}{2})^2}$ (2.17)

for all pairs (r_i, ℓ_i) such that either $r_i = q_i$ and $\ell_i = 0$, or $r_i \in \{1, \dots, q_i\}$, $\ell_i \in \{1, \dots, \min(r_i, q_i - 1)\}$ and r_i and ℓ_i not equal to $q_i/2$ at the same time. For $i \in \{d_1 + 1, \dots, d_1 + d_2\}$ suppose that

$$\lim_{n \rightarrow \infty} \|f_n^{(i)} \star_{r_i}^{\ell_i} f_n^{(i)}\| = 0 \quad (2.18)$$

for all $r_i \in \{1, \dots, q_i\}$ and $\ell_i \in \{1, \dots, \min(r_i, q_i - 1)\}$. Then

$$(I_{q_1}(f_n^{(1)}), \dots, I_{q_{d_1+d_2}}(f_n^{(d_1+d_2)}))$$

converges in distribution to $(G_1, \dots, G_{d_1}, N_{d_1+1}, \dots, N_{d_1+d_2})$ as $n \rightarrow \infty$.

Example 2.22. We illustrate Theorem 2.21 with an example related to the theory of random graphs; the reader is referred to Penrose (2003) for an introduction to this topic. Note that we will allow the underlying Poisson measure to depend on n ; see Remark 2.19. Let $d \geq 1$, and define \mathbf{Y} and η_n as in Section 2.4. Let $\{r_n : n \geq 1\}$ be a sequence of strictly positive numbers decreasing to zero. For every n , we define $D_n := (V_n, E_n)$ to be the random ‘disk graph’ obtained as follows: $V_n = \{Y_i : i = 1, \dots, N(n)\}$ and two vertices $Y_i, Y_j \in V_n$ are connected by an edge if and only if their Euclidean distance is strictly positive and less than r_n (in particular, D_n has no loops). Now let Λ be a feasible connected graph (in the sense of Bourguin and Peccati (2012); Lachièze-Rey and Peccati (2013b); Penrose (2003)) with q vertices, where $q \neq 2, 4$. For every n , we define L_n to be the random variable equal to the number of induced subgraphs of D_n that are isomorphic to Λ , that is, L_n is equal to the number of subsets of the type $Y_{(q)} = \{Y_{i_1}, \dots, Y_{i_q}\} \subset V_n$ such that the restriction of D_n to $Y_{(q)}$ is isomorphic to Λ . We also set $\tilde{L}_n = (L_n - \mathbb{E}[L_n]) / \text{Var}(L_n)^{1/2}$. Now assume that $nr_n^d \rightarrow 0$ and $n^q (r_n^d)^{q-1} \rightarrow \infty$. According to the discussion contained e.g. in Penrose (2003, Chapter 3) or Lachièze-Rey and Peccati (2013b, Section 3), one has that the following four facts are in order: (i) $\mathbb{E}[L_n] \approx K_0 n^q (r_n^d)^{q-1}$ (for some finite constant $K_0 > 0$), (ii) $\text{Var}(L_n) \approx K_1 n^q (r_n^d)^{q-1}$ (for some finite constant $K_1 > 0$), (iii) there exists a sequence of multiple integrals of order q with respect to $\hat{\eta}_n = \eta_n - \mu_n$, say $I_q(f_n)$, such that, as $n \rightarrow \infty$, $\mathbb{E}[(\tilde{L}_n - I_q(f_n))^2] \rightarrow 0$, and (iv) the kernels

$\{f_n\}$ verify the asymptotic relation (2.18) (where, for every n , the contractions and norms have to be considered with respect to the measure μ_n), so that \tilde{L}_n converges in distribution to a standard Gaussian random variable as $n \rightarrow \infty$. Now consider a sequence $\{F'_n\}$ of degenerate U -statistics of order 2 as in Theorem 2.17 (for instance, those appearing in Example 2.18). Since each F'_n is a double Wiener-Itô integral (with respect to η_n) verifying condition (2.17) and $q \neq 4$, we can directly apply Theorem 2.21 in the case $d_1 = d_2 = 1$, $q_1 = 2$ and $q_2 = q$, and conclude that, as $n \rightarrow \infty$, the pair (F'_n, \tilde{L}_n) converges in distribution to a vector (G, N) composed of independent random variables such that G has distribution $\bar{\Gamma}_\nu$ and N follows a standard Gaussian random variable.

We finally show how one can use the results of the present paper to deal with a hybrid Poisson/Gamma convergence (we just consider two-dimensional vectors in order to simplify the discussion, but there is no additional difficulty in considering vectors of higher dimensions). Let $\nu, \lambda > 0$ and let (G, P) be a vector consisting of independent random variables such that G has distribution $\bar{\Gamma}_\nu$ and P has a Poisson distribution with mean λ . We fix an even integer $q \geq 2$, and consider a sequence $\{f_n : n \geq 1\}$ of kernels with $f_n \in L^2_{\text{sym}}(\mu_n^q)$ and such that the technical conditions stated in Section 3-(VIII) are satisfied. We also consider a sequence $\{H_n : n \geq 1\}$ of random variables such that: (a) each H_n is a functional of the Poisson measure η , which is in the domain of the Malliavin derivative D and takes values in $\mathbb{Z}_+ = \{0, 1, 2, \dots\}$, (b) the numerical sequence

$$n \mapsto \mathbb{E} \int_{\mathcal{Z}} (D_z H_n)^2 \mu_n(dz), \quad n \geq 1, \tag{2.19}$$

is bounded.

Theorem 2.23 (Gamma/Poisson hybrid convergence). *Assume that*

$$\lim_{n \rightarrow \infty} q! \|f_n\|^2 = 2\nu$$

and that

$$\begin{aligned} \lim_{n \rightarrow \infty} \|f_n \star_r^\ell f_n\| &= 0 \\ \text{and } \lim_{n \rightarrow \infty} \|f_n \tilde{\star}_{q/2}^{q/2} f_n - c_q f_n\| &= 0 \quad \text{with } c_q = \frac{4}{\left(\frac{q}{2}\right)! \left(\frac{q}{2}\right)^2} \end{aligned} \tag{2.20}$$

for all pairs (r, ℓ) such that either $r = q$ and $\ell = 0$, or $r \in \{1, \dots, q\}$, $\ell \in \{1, \dots, \min(r, q - 1)\}$ and r and ℓ not equal to $q/2$ at the same time. We also assume that, as $n \rightarrow \infty$,

$$\begin{aligned} \mathbb{E}[H_n] &\rightarrow \lambda, & \mathbb{E}|\lambda - \langle DH_n, -DL^{-1}H_n \rangle| &\rightarrow 0, \\ \text{and } \mathbb{E} \int_{\mathcal{Z}} |D_z H_n (D_z H_n - 1) D_z L^{-1} H_n| \mu_n(dz) &\rightarrow 0. \end{aligned} \tag{2.21}$$

Then $(I_q(f_n), H_n)$ converges in distribution to (G, P) as $n \rightarrow \infty$.

Example 2.24. We consider the same framework and notation as in Example 2.22. Here, we take $q \geq 2$ to be a general integer (which can be possibly equal to 2 or 4), whereas Λ is a feasible connected graph of order q . We stress that, for every n , the random variable L_n is a functional of the Poisson measure η_n on \mathbb{R}^d , whose control measure is given by $\mu_n(dx) = np(x)dx$. We put $r_n = n^{-dq/(q-1)}$, in such a way that $n^q (r_n^d)^{q-1} = 1$. According e.g. to the analysis contained in

Penrose (2003, Chapter 3) or Bourguin and Peccati (2012, Section 2.4), one has that the following two facts are in order: (i) there exists a constant $\lambda > 0$ such that $\mathbb{E}[L_n] \approx \text{Var}(L_n) \approx \lambda n^q (r_n^d)^{q-1} = \lambda$, and (ii) the sequence $H_n = L_n$ satisfies (2.19), as well as the asymptotic relations (2.21) (here, for every n , the Malliavin operators are defined with respect to the random measure η_n and the inner products and integrals are obtained by integrating with respect to $\mu = \mu_n$), so that L_n converges in distribution to a Poisson random variable with mean λ . Considering a sequence $\{F'_n\}$ of degenerate U -statistics of order 2 as in Theorem 2.17 (see e.g. Example 2.18), one has that each F'_n is a double integral verifying condition (2.20). We can therefore apply Theorem 2.23 and infer that, as $n \rightarrow \infty$, the pair (F'_n, L_n) converges in distribution to a vector (G, P) composed of independent random variables such that G is distributed according to $\bar{\Gamma}_\nu$ and P has a Poisson distribution with mean λ .

Example 2.25. Let us consider a Poisson measure η_n of k -dimensional flats in \mathbb{R}^d with $2k < d$ (where the flats are suitably parameterized to fit into our framework). We assume that the distribution of η_n is invariant under rigid motions for each n , and that η_n has intensity $n \geq 1$. Let us fix a closed convex set $W \subset \mathbb{R}^d$ with volume one and define the distance $\text{dist}_W(E, F)$ of two k -flat E, F as the minimum over the Euclidean distances of $x_E \in E \cap W$ and $x_F \in F \cap W$. By M_n we denote the number of pairs (E, F) of distinct flats of η_n such that $\text{dist}_W(E, F) \leq r_n$, where $r_n = n^{-2/(d-2k)}$. According to Theorem 2.1 in Schulte and Thäle (2012) we know that (i) there exists a constant $0 < \lambda < \infty$ (depending on d, k and W) such that $\mathbb{E}[M_n] \approx \mathbb{V}M_n \approx \lambda$, (ii) the asymptotic relations (2.21) are satisfied, and (iii) M_n fulfills the technical condition (2.19). Thus, M_n converges in distribution to a Poisson random variable with mean λ . Let now $\{F'_n\}$ be a sequence of degenerate U -statistics of order 2 as in Theorem 2.17; see Example 2.18. Then, as in the previous example, each F'_n is a double integral such that condition (2.20) is verified. Thus, Theorem 2.23 can be applied to show that the random vector (F'_n, M_n) converges in distribution to a random vector (G, P) with independent components such that G has distribution $\bar{\Gamma}_\nu$ and P has a Poisson distribution with mean λ .

Example 2.26. Let $W \subset \mathbb{R}^d$ be a closed convex set with volume one and let \mathbf{Y} be a sequence of i.i.d. points in W , which are uniformly distributed and whose random number $N(n)$ follows a Poisson distribution with parameter $n \in \mathbb{N}$. Any $d + 1$ distinct points of \mathbf{Y} form a non-degenerate random simplex in W . Define $r_n := n^{-(d+1)}$ and let V_n be the total number of such simplices whose volume does not exceed r_n . Then (i) there exists $0 < \lambda < \infty$ (depending on d and W) such that $\mathbb{E}[V_n] \approx \mathbb{V}V_n \approx \lambda$, (ii) the asymptotic relations (2.21) are satisfied, and (iii) V_n fulfills the technical condition (2.19) so that the law of V_n converges, as $n \rightarrow \infty$ to a Poisson distribution with mean λ . This can be seen from Theorem 2.5 in Schulte and Thäle (2012). Define the sequence $\{F'_n\}$ of degenerate U -statistics as in Example 2.18 or, more generally, as in Theorem 2.17. Then, following the same line of reasoning as above, Theorem 2.23 can be applied to show that the random vector (F'_n, V_n) converges in distribution to a random vector (G, P) with independent components such that G has distribution $\bar{\Gamma}_1$ and P has a Poisson distribution with mean λ .

3. Background material

In this section we collect definitions and results that are needed in the statements and proofs of our results. For more details, we refer to the monographs [Peccati and Taqqu \(2011\)](#); [Privault \(2009\)](#) or to the papers [Last and Penrose \(2011\)](#); [Nualart and Vives \(1990\)](#).

(I) Poisson measures. We shall denote by η a Poisson measure with non-atomic and σ -finite control measure μ on some Polish space \mathcal{Z} (which is endowed with the Borel σ -field \mathcal{Z}). Recall that η is a collection $\{\eta(B) : B \in \mathcal{X}_0\}$ of random variables indexed by the members of $\mathcal{X}_0 = \{B \in \mathcal{Z} : \mu(B) < \infty\}$ such that: (a) $\eta(B)$ follows a Poisson distribution with mean $\mu(B)$ for all $B \in \mathcal{X}_0$, and (b) whenever $A, B \in \mathcal{X}_0$ are disjoint, $\eta(A)$ and $\eta(B)$ are independent random variables. By P_η we will denote the distribution of η (on the space of σ -finite counting measures on \mathcal{Z}).

(II) L^2 -spaces. For $q \geq 1$ we denote by $L^2(\mu^q)$ the L^2 -space $L^2(\mathcal{Z}^q, \mathcal{Z}^{\otimes q}, \mu^q)$ and by $L^2_{\text{sym}}(\mu^q)$ the subspace of $L^2(\mu^q)$ consisting of functions that are μ^q -a.e. invariant under permutations of its arguments, so called symmetric functions. Suppressing the dependency on q , the scalar product and the norm in $L^2(\mu^q)$ (and $L^2_{\text{sym}}(\mu^q)$) are denoted by $\langle \cdot, \cdot \rangle$ and $\| \cdot \|$, respectively. In addition, we let $L^2(P_\eta)$ be the space of square-integrable functionals of η . To avoid confusion we will use capitals to indicate elements of $L^2(P_\eta)$ and lower cases for elements of $L^2(\mu^q)$ or $L^2_{\text{sym}}(\mu^q)$. We finally introduce the space $L^2(\mathbb{P}, L^2(\mu)) = L^2(\Omega \times \mathcal{Z}, \mathcal{F} \otimes \mathcal{Z}, \mathbb{P} \otimes \mu)$ as the space of jointly measurable mappings $u : \Omega \times \mathcal{Z} \rightarrow \mathbb{R}$ such that $\mathbb{E} \int_{\mathcal{Z}} u(z)^2 \mu(dz) < \infty$ (recall that $(\Omega, \mathcal{F}, \mathbb{P})$ is the underlying probability space).

(III) Multiple stochastic integrals. For every integer $q \geq 1$ and every deterministic function $f \in L^2_{\text{sym}}(\mu^q)$ let us indicate by $I_q(f)$ the *multiple Wiener-Itô stochastic integral* of order q of f with respect to the compensated Poisson measure $\eta - \mu$. For general $f \in L^2(\mu^q)$ we put $I_q(f) := I_q(\tilde{f})$, where $\tilde{f}(x_1, \dots, x_q) = (q!)^{-1} \sum_{\pi} f(x_{\pi(1)}, \dots, x_{\pi(q)})$ is the canonical symmetrization of f and the sum in its definition runs over all $q!$ permutations π of $\{1, \dots, q\}$. The multiple stochastic integrals satisfy the following properties:

$$\mathbb{E}[I_{q_i}(f_i)] = 0 \quad (i = 1, 2), \quad \text{and} \quad \mathbb{E}[I_{q_1}(f_1)I_{q_2}(f_2)] = q_1! \langle f_1, f_2 \rangle \mathbf{1}_{(q_1 = q_2)}$$

for any $q_1, q_2 \in \{1, 2, \dots\}$, $f_1 \in L^2_{\text{sym}}(\mu^{q_1})$ and $f_2 \in L^2_{\text{sym}}(\mu^{q_2})$.

(IV) Chaotic representation property. The q -th Wiener chaos W_q associated with the Poisson measure η is the Hilbert space $W_q = \{I_q(f) : f \in L^2_{\text{sym}}(\mu^q)\}$. In addition, we put $W_0 := \mathbb{R}$. It is a crucial property of η that $L^2(P_\eta)$ can be written as a direct sum of Wiener chaoses, i.e. $L^2(P_\eta) = \bigoplus_{q=0}^{\infty} W_q$. As a consequence, every $F \in L^2(P_\eta)$ admits a (unique) chaotic decomposition in the sense that

$$F = \mathbb{E}[F] + \sum_{q=1}^{\infty} I_q(f_q) \tag{3.1}$$

with suitable functions $f_q \in L^2_{\text{sym}}(\mu^q)$ and where the series converges in $L^2(P_\eta)$.

(V) The Malliavin operators D , L^{-1} and δ . The domain $\text{dom } D$ of the *derivative operator* D is the set of all $F \in L^2(P_\eta)$ admitting a chaos decomposition

(3.1) such that $\sum_{q=1}^{\infty} q q! \|f_q\|^2 < \infty$. For such F the random function $\mathcal{Z} \ni z \mapsto D_z F$ is defined by

$$D_z F = \sum_{q=1}^{\infty} q I_{q-1}(f_q(z, \cdot)), \quad (3.2)$$

where $f_q(z, \cdot)$ is the function f_q with one of its argument fixed to be z . Notice that $DF \in L^2(\mathbb{P}, L^2(\mu))$. The derivative operator can be also characterized as an ‘‘add-one cost operator’’, as follows; see Last and Penrose (2011); Nualart and Vives (1990) for proofs of this fact. For $F \in L^2(P_\eta)$ and $z \in \mathcal{Z}$, let $F_z(\eta)$ be the random variable $F(\eta + \delta_z)$. Then, for $F \in \text{dom } D$ and μ -almost every $z \in \mathcal{Z}$, we have the identity $D_z F = F_z - F$, a.s.- \mathbb{P} . Throughout the text, we also implicitly use the following converse statement (the proof is an elementary consequence of the main findings of Last and Penrose (2011), and is included for the sake of completeness).

Lemma 3.1. *Let $F \in L^2(P_\eta)$ be such that $\mathbb{E} \int_{\mathcal{Z}} (F_z - F)^2 \mu(dz) < \infty$. Then, $F \in \text{dom } D$.*

Proof: For every $z \in \mathcal{Z}$, define the ‘trajectorial’ difference operator $D'_z F(\eta) = F_z(\eta) - F(\eta)$. According to Last and Penrose (2011, Theorem 1.3), the square-integrable random variable F admits a chaotic decomposition of the type (3.1), with

$$f_q(z_1, \dots, z_q) = \frac{1}{q!} \mathbb{E}[D'_{z_1} \dots D'_{z_q} F], \quad q = 1, 2, \dots$$

(in particular, the deterministic function on the right-hand side of the previous equation is a well-defined element of $L^2_{\text{sym}}(\mu^q)$ for every $F \in L^2(P_\eta)$ and every $q \geq 1$). (An alternate prove of this result could be deduced from Ito (1988, Equation (7.4)) after an adequate translation of the necessary notions from the language of white noise analysis into the formalism of the present paper.) In view of the assumptions, there exists a measurable set \mathcal{Z}' such that $\mu(\mathcal{Z} \setminus \mathcal{Z}') = 0$ and $\mathbb{E}[(D'_z F)^2] = \mathbb{E}[(F_z - F)^2] < \infty$ for every $z \in \mathcal{Z}'$. It follows that the statement is proved once we show that, for every $z \in \mathcal{Z}'$, the chaotic decomposition of $D'_z F$ coincides with the right-hand side of (3.2). Again by virtue of Last and Penrose (2011, Theorem 1.3), one has that the q th integrand in the chaotic decomposition of $D'_z F$ is given by the mapping

$$(z_1, \dots, z_q) \mapsto \frac{1}{q!} \mathbb{E}[D'_{z_1} \dots D'_{z_q} D'_z F] = (q+1) f_{q+1}(z, z_1, \dots, z_q),$$

which yields the desired conclusion. \square

For any $F \in L^2(P_\eta)$ with chaotic decomposition (3.1) satisfying $\mathbb{E}[F] = 0$ we put

$$L^{-1} F = - \sum_{q=1}^{\infty} q^{-1} I_q(f_q).$$

The operator L^{-1} is the so-called *pseudo-inverse of the Ornstein-Uhlenbeck generator*. Finally, we observe that, due to the chaotic representation property of $\hat{\eta}$, every random function $u \in L^2(\mathbb{P}, L^2(\mu))$ admits a (unique) representation of the type

$$u_z = \sum_{q=0}^{\infty} I_q(f_q(z, \cdot)), \quad z \in \mathcal{Z}, \quad (3.3)$$

where, for every z , the kernel $f_q(z, \cdot)$ is an element of $L^2_{\text{sym}}(\mu^q)$. The domain of the *divergence operator*, denoted by $\text{dom } \delta$, is defined as the collections of those $u \in L^2(\mathbb{P}, L^2(\mu))$ such that the chaotic expansion (3.3) verifies the condition

$$\sum_{q=0}^{\infty} (q+1)! \|f_q\|^2 < \infty.$$

If $u \in \text{dom } \delta$, then the random variable $\delta(u)$ is defined as

$$\delta(u) = \sum_{q=0}^{\infty} I_{q+1}(\tilde{f}_q),$$

where \tilde{f}_q stands for the canonical symmetrization of f_q (as a function in $q+1$ variables). The following classic result, proved e.g. in [Nualart and Vives \(1990\)](#), yields a characterization of δ as the adjoint of the derivative D .

Lemma 3.2 (Integration by parts formula). *For every $G \in \text{dom } D$ and every $u \in \text{dom } \delta$, one has that*

$$\mathbb{E}[G\delta(u)] = \mathbb{E}[\langle DG, u \rangle], \tag{3.4}$$

where, more explicitly,

$$\langle DG, u \rangle = \int_{\mathcal{Z}} D_z G \times u(z) \mu(dz).$$

(VI) Contractions. Let $f \in L^2_{\text{sym}}(\mu^q)$ for some integer $q \geq 1$ and $r \in \{0, \dots, p\}$, $\ell \in \{1, \dots, r\}$. The contraction kernel $f \star_r^\ell f$ on $\mathcal{Z}^{2q-r-\ell}$ acts on the tensor product $f \otimes f$ first by identifying r variables and then integrating out ℓ among them. More formally,

$$\begin{aligned} f \star_r^\ell f(\gamma_1, \dots, \gamma_{r-\ell}, t_1, \dots, t_{q-r}, s_1, \dots, s_{q-r}) \\ = \int_{\mathcal{Z}^\ell} f(z_1, \dots, z_\ell, \gamma_1, \dots, \gamma_{r-\ell}, t_1, \dots, t_{q-r}) \\ \times f(z_1, \dots, z_\ell, \gamma_1, \dots, \gamma_{r-\ell}, s_1, \dots, s_{q-r}) \mu^\ell(d(z_1, \dots, z_\ell)). \end{aligned}$$

In addition, we put

$$\begin{aligned} f \star_r^0 f(\gamma_1, \dots, \gamma_r, t_1, \dots, t_{q-r}, s_1, \dots, s_{q-r}) \\ = f(\gamma_1, \dots, \gamma_r, t_1, \dots, t_{q-r}) f(\gamma_1, \dots, \gamma_r, s_1, \dots, s_{q-r}). \end{aligned}$$

Besides the contraction $f \star_r^\ell f$, we will also deal with its canonical symmetrization $f \tilde{\star}_r^\ell f$, which is defined as

$$(f \tilde{\star}_r^\ell f)(x_1, \dots, x_{2q-r-\ell}) = \frac{1}{(2q-r-\ell)!} \sum_{\pi} (f \star_r^\ell f)(x_{\pi(1)}, \dots, x_{\pi(2q-r-\ell)}),$$

where the sum runs over all $(2q-r-\ell)!$ permutations of the set $\{1, \dots, 2q-r-\ell\}$.

(VII) Product formula. Let $q_1, q_2 \geq 1$ be integers, $f_1 \in L^2_{\text{sym}}(\mu^{q_1})$ and $f_2 \in L^2_{\text{sym}}(\mu^{q_2})$ be as in the previous paragraph. In terms of the contractions of f_1 and f_2 one can express the product of $I_{q_1}(f_1)$ and $I_{q_2}(f_2)$ as follows:

$$I_{q_1}(f_1) I_{q_2}(f_2) = \sum_{r=0}^{\min(q_1, q_2)} r! \binom{q_1}{r} \binom{q_2}{r} \sum_{\ell=0}^r \binom{r}{\ell} I_{q_1+q_2-r-\ell}(f_1 \tilde{\star}_r^\ell f_2); \tag{3.5}$$

see [Peccati and Taqqu \(2011, Proposition 6.5.1\)](#). In the particular case $q_1 = q_2 =: q$ and $f_1 = f_2$, we may define $G_0^q f := q! \|f\|^2$ and

$$G_p^q f := \sum_{r=0}^q \sum_{\ell=0}^r \mathbf{1}(2q - r - \ell = p) r! \binom{q}{r}^2 \binom{r}{\ell} f \star_r^\ell f$$

for $p \in \{1, \dots, 2q\}$, which allows us to re-write (3.5) in the more compact form

$$I_q^2(f) = \sum_{p=0}^{2q} I_p(G_p^q f). \quad (3.6)$$

(VIII) Technical assumptions. Whenever we deal with a multiple stochastic integral or a sequence $I_q(f_n)$ of such integrals with $f_n \in L_{\text{sym}}^2(\mu_n^q)$ we will (implicitly) assume that the following technical conditions are satisfied:

- i) for any $r \in \{1, \dots, q\}$, the contraction $f_n \star_r^{q-r} f_n$ is an element of $L^2(\mu_n^r)$;
- ii) for any $r \in \{1, \dots, q\}$, $\ell \in \{1, \dots, r\}$ and $(z_1, \dots, z_{2q-r-\ell}) \in \mathcal{Z}^{2q-r-\ell}$ we have that $(|f_n| \star_r^\ell |f_n|)(z_1, \dots, z_{2q-r-\ell})$ is well defined and finite;
- iii) for any $k \in \{0, \dots, 2(q-1)\}$ and any r and ℓ satisfying $k = 2(q-1) - r - \ell$ we have that

$$\int_{\mathcal{Z}} \sqrt{\int_{\mathcal{Z}} (f_n(z, \cdot) \star_r^\ell f_n(z, \cdot))^2 d\mu_n^k \mu_n(dz)} < \infty.$$

We remark that (iii) is automatically satisfied if the control measure μ is finite (which is the case in our Examples 2.22 and 2.24-2.26). Intuitively, conditions (i)-(iii) ensure that every manipulation involving contraction kernels performed below is justified and is in fact valid. For the detailed role of these conditions and their implications we refer to [Lachièze-Rey and Peccati \(2013a\)](#) or [Peccati et al. \(2010\)](#).

4. Proofs of the results

4.1. *Proof of Theorem 2.1.* Before entering the details of the proof of Theorem 2.1 we recall some facts related to Stein's method for the Gamma distribution established by [Luk \(1994\)](#); see also [Pickett \(2004\)](#) for refinements in the case of an integer-valued parameter ν . We start by considering the *second-order* Stein equation

$$h(x - \nu) - \mathbb{E}[h(G^*(\nu))] = 2xg''(x) - (x - \nu)g'(x), \quad x > 0, \quad (4.1)$$

where $G^*(\nu) = G(\nu) + \nu$, with $G(\nu)$ distributed according to $\bar{\Gamma}_\nu$, and $h \in \mathcal{H}^3$. It is shown in [Luk \(1994, Theorem 1\)](#) that (4.1) admits a solution V_h such that $\|V_h^{(j)}\|_\infty \leq \frac{2}{j} \|h^{(j)}\|_\infty$ for $j = 1, 2, 3$. Note that the assumption $h \in \mathcal{H}^3$ automatically yields that h has sub-exponential growth, so that [Luk \(1994, Theorem 1\)](#) can directly be applied.

Now, we turn to the *first-order* Stein operator T for $\bar{\Gamma}_\nu$, which acts on differentiable functions $f : \mathbb{R} \rightarrow \mathbb{R}$. It is given by

$$Tf(x) = 2(x + \nu)_+ f'(x) - xf(x), \quad x \in \mathbb{R}.$$

The associated first-order Stein equation is

$$h(x) - \mathbb{E}[h(G(\nu))] = Tf(x), \quad x \in \mathbb{R},$$

where $h \in \mathcal{H}^3$. For such an h , a solution U_h of the Stein equation – in what follows, sometimes called a *Stein solution* – is provided by

$$U_h(x) = \begin{cases} -\frac{1}{x}(h(x) - \mathbb{E}[h(G(\nu))]) & : x \leq -\nu \\ V'_h(x + \nu) & : x > -\nu. \end{cases} \quad (4.2)$$

Recall that the probability density of $G(\nu)$ is given by

$$g_\nu(x) = \frac{2^{-\nu/2}}{\Gamma(\nu/2)}(x + \nu)^{\nu/2-1} e^{-(x+\nu)/2} \mathbf{1}_{\{x > -\nu\}}.$$

Since for our choice of the test function h the mapping $x \mapsto U_h(x) = V'_h(x + \nu)$ is bounded on (ν, ∞) , we can use [Stein \(1986, Lemma 4\)](#) to deduce that, necessarily,

$$U_h(x) = \frac{1}{2(x + \nu)g_\nu(x)} \int_{-\nu}^x (h(y) - \mathbb{E}[h(G(\nu))]) g_\nu(y) dy, \quad x > -\nu,$$

yielding that $x \mapsto U_h(x)$ is continuous on \mathbb{R} , as deduced from a simple application of de l'Hôpital's rule at $x = -\nu$. Also, U_h is twice differentiable on $\mathbb{R} \setminus \{-\nu\}$ and satisfies the estimates $\|U_h\|_\infty \leq \max(2, 2/\nu) =: c_0$, $\|U'_h\|_\infty \leq \max(1, 1/\nu + 1/\nu^2) = c_1$ and $\|U''_h\|_\infty \leq \max(2/3, 2/(3\nu) - 3/\nu^2 + 4/\nu^3) = c_2$ (here, c_1 and c_2 are the constants from [Theorem 2.1](#)). We stress that, albeit $U_h(x)$ is continuous on \mathbb{R} , such a function is in general not differentiable at $x = -\nu$ (it is however right- and left-differentiable at such a point). We remark that the quantities 2, 1 and $2/3$ appearing in the constants c_0, c_1 and c_2 come from smoothness estimates for [\(4.2\)](#) on the interval $(-\nu, \infty)$, whereas the presence of the constants $2/\nu, 1/\nu + 1/\nu^2$ and $2/(3\nu) - 3/\nu^2 + 4/\nu^3$ is explained by elementary estimates of [\(4.2\)](#) on the interval $(-\infty, -\nu]$.

Let now \mathcal{F}^2 be the space of continuous functions f on \mathbb{R} , which are twice differentiable on $\mathbb{R} \setminus \{-\nu\}$ and satisfy

$$\|f\|_\infty \leq c_0, \quad \|f'\|_\infty \leq c_1 \quad \text{and} \quad \|f''\|_\infty \leq c_2.$$

In the light of the previous discussion, we conclude that

$$d_3(F, \bar{\Gamma}_\nu) \leq \sup_{f \in \mathcal{F}^2} |\mathbb{E}[2(F + \nu)_+ f'(F) - Ff(F)]|, \quad (4.3)$$

where F and ν are as in the statement of [Theorem 2.1](#) and where here and below $f'(-\nu)$ stands for the left-sided derivative of f at $-\nu$, i.e. $f'(-\nu) = \lim_{x \nearrow -\nu} f'(x)$.

We also refer the reader to [Lemma 1.3 in Nourdin and Peccati \(2009b\)](#) and the references cited therein. The estimate [\(4.3\)](#) is the starting point of the proof of [Theorem 2.1](#).

Proof of Theorem 2.1: We have to show that the right-hand side of [\(4.3\)](#) is bounded from above by the right-hand side of [\(2.1\)](#). This is done by borrowing some ideas from [Schulte \(2012\)](#). To start with, consider $f \in \mathcal{F}^2$, write $F(\eta)$ instead of F to emphasize the dependency of F on η and fix $z \in \mathcal{Z}$. Because of the non-differentiability of the Stein-solution at $-\nu$, we will have to distinguish the three cases a) $F(\eta) \leq -\nu$, $F(\eta + \delta_z) \leq -\nu$ or $F(\eta) > -\nu$, $F(\eta + \delta_z) > -\nu$, b) $F(\eta) \leq -\nu < F(\eta + \delta_z)$ and c) $F(\eta + \delta_z) \leq -\nu < F(\eta)$. For a) we use a Taylor expansion

to see that

$$\begin{aligned} D_z f(F(\eta)) &= f(F(\eta + \delta_z)) - f(F(\eta)) \\ &= f'(F(\eta))(F(\eta + \delta_z) - F(\eta)) + R(F(\eta + \delta_z) - F(\eta)) \\ &= f'(F(\eta))D_z F(\eta) + R_a(D_z F(\eta)), \end{aligned}$$

where the reminder R_a is such that $|R_a(x)| \leq \frac{1}{2}\|f''\|_\infty x^2 = \frac{1}{2}c_2 x^2$; recall that f is differentiable on $\mathbb{R} \setminus \{-\nu\}$, as well as right- and left-differentiable at $x = -\nu$. For case b) we also use a Taylor expansion to see that

$$\begin{aligned} D_z f(F(\eta)) &= f(F(\eta + \delta_z)) - f(F(\eta)) \\ &= f(F(\eta + \delta_z)) - f(-\nu) + f(-\nu) - f(F(\eta)) \\ &= f'(-\nu+)(F(\eta + \delta_z) + \nu) + \frac{1}{2}f''(\tilde{F}(\eta))(F(\eta + \delta_z) + \nu)^2 \\ &\quad + f'(F(\eta))(-\nu - F(\eta)) + \frac{1}{2}f''(\hat{F}(\eta))(-\nu - F(\eta))^2 \\ &= f'(F(\eta))D_z F(\eta) - f'(F(\eta))(F(\eta + \delta_z) + \nu) \\ &\quad + f'(-\nu+)(F(\eta + \delta_z) + \nu) + \frac{1}{2}f''(\tilde{F}(\eta))(F(\eta + \delta_z) + \nu)^2 \\ &\quad + \frac{1}{2}f''(\hat{F}(\eta))(F(\eta) + \nu)^2 \\ &=: f'(F(\eta))D_z F(\eta) + R_b(F(\eta), z, \nu) \end{aligned}$$

with some $\tilde{F} \in (-\nu, F(\eta + \delta_z))$, $\hat{F} \in (F(\eta), -\nu)$ and where $f'(-\nu+)$ stands for the right-sided derivative of f at $-\nu$. Similarly, in case c) we find that

$$\begin{aligned} D_z f(F(\eta)) &= f(F(\eta + \delta_z)) - f(F(\eta)) \\ &= f(F(\eta + \delta_z)) - f(-\nu) + f(-\nu) - f(F(\eta)) \\ &= f'(-\nu-)(F(\eta + \delta_z) + \nu) + \frac{1}{2}f''(\tilde{F}(\eta))(F(\eta + \delta_z) + \nu)^2 \\ &\quad + f'(F(\eta))(-\nu - F(\eta)) + \frac{1}{2}f''(\hat{F}(\eta))(-\nu - F(\eta))^2 \\ &= f'(F(\eta))D_z F(\eta) - f'(F(\eta))(F(\eta + \delta_z) + \nu) \\ &\quad + f'(-\nu-)(F(\eta + \delta_z) + \nu) + \frac{1}{2}f''(\tilde{F}(\eta))(F(\eta + \delta_z) + \nu)^2 \\ &\quad + \frac{1}{2}f''(\hat{F}(\eta))(F(\eta) + \nu)^2 \\ &=: f'(F(\eta))D_z F(\eta) + R_c(F(\eta), z, \nu) \end{aligned}$$

again with some $\tilde{F} \in (F(\eta + \delta_z), -\nu)$, $\hat{F} \in (-\nu, F(\eta))$ and where $f'(-\nu-)$ stands for the left-sided derivative of f at ν . Summarizing, we conclude that

$$D_z f(F(\eta)) = f'(F(\eta))D_z F(\eta) + R(F(\eta), z, \nu) \quad (4.4)$$

(recall that $f'(-\nu) = f'(-\nu-)$ by convention), where the global reminder term $R(F(\eta), z, \nu)$ is given by

$$\begin{aligned} R(F(\eta), z, \nu) &= R_a(F(\eta))(\mathbf{1}_{\{F(\eta), F(\eta + \delta_z) > -\nu\}} + \mathbf{1}_{\{F(\eta), F(\eta + \delta_z) \leq -\nu\}}) \\ &\quad + R_b(F(\eta), z, \nu) \mathbf{1}_{\{F(\eta) \leq -\nu < F(\eta + \delta_z)\}} \\ &\quad + R_c(F(\eta), z, \nu) \mathbf{1}_{\{F(\eta + \delta_z) \leq -\nu < F(\eta)\}}. \end{aligned}$$

We have seen that R_a has the property that $|R_a(x)| \leq \frac{1}{2}\|f''\|_\infty x^2 = \frac{1}{2}c_2x^2$. For R_b and R_c we notice that in these cases $|F(\eta + \delta_z) + \nu| \leq |D_zF(\eta)|$ and $|F(\eta) + \nu| \leq |D_zF(\eta)|$, which together with the properties of $f \in \mathcal{F}^2$ leads to the bound

$$\begin{aligned} & |R(F(\eta), z, \nu)| \\ & \leq \frac{1}{2}c_2|D_zF(\eta)|^2(\mathbf{1}_{\{F(\eta), F(\eta+\delta_z) > -\nu\}} + \mathbf{1}_{\{F(\eta), F(\eta+\delta_z) \leq -\nu\}}) \\ & \quad + (2c_1|D_zF(\eta)| + c_2|D_zF(\eta)|^2)\mathbf{1}_{\{F(\eta) \leq -\nu < F(\eta+\delta_z)\}} \\ & \quad + (2c_1|D_zF(\eta)| + c_2|D_zF(\eta)|^2)\mathbf{1}_{\{F(\eta+\delta_z) \leq -\nu < F(\eta)\}} \\ & \leq c_2|D_zF(\eta)|^2 + 2c_1|D_zF(\eta)|(\mathbf{1}_{\{F(\eta) \leq -\nu < F(\eta+\delta_z)\}} + \mathbf{1}_{\{F(\eta+\delta_z) \leq -\nu < F(\eta)\}}) \\ & = c_2|D_zF(\eta)|^2 + 2c_1(D_z\mathbf{1}_{\{F(\eta) > -\nu\}})(D_zF(\eta)). \end{aligned}$$

Using now the integration by parts formula from Malliavin calculus, (3.4) in Lemma 3.2, and simplifying the resulting expression we find

$$\mathbb{E}[Ff(F)] = \mathbb{E}[LL^{-1}Ff(F)] = \mathbb{E}[-\delta(DL^{-1}F)f(F)] = \mathbb{E}[\langle Df(F), -DL^{-1}F \rangle],$$

which in view of (4.4) leads to

$$\mathbb{E}[\langle Df(F), -DL^{-1}F \rangle] = \mathbb{E}[f'(F)\langle DF, -DL^{-1}F \rangle] + \mathbb{E}[\langle R(F, z, \nu), -DL^{-1}F \rangle].$$

Consequently, because of the above estimate on $|R(F(\eta), z, \nu)|$,

$$\begin{aligned} & |\mathbb{E}[2(F + \nu)_+ f'(F) - Ff(F)]| \\ & \leq |\mathbb{E}[f'(F)(2(F + \nu)_+ - \langle DF, -DL^{-1}F \rangle)]| + |\mathbb{E}[\langle R(F, z, \nu), -DL^{-1}F \rangle]| \\ & \leq c_1\mathbb{E}|2(F + \nu)_+ - \langle DF, -DL^{-1}F \rangle| + c_2 \int_{\mathcal{Z}} \mathbb{E}[|D_zF|^2|DL^{-1}F|] \mu(dz) \\ & \quad + 2c_1 \int_{\mathcal{Z}} \mathbb{E}[(D_z\mathbf{1}_{\{F > -\nu\}})(D_zF)|D_zL^{-1}F|] \mu(dz). \end{aligned}$$

This shows the first inequality (2.1) in Theorem 2.1. The second estimate (2.2) follows from (2.1) and the assumption that $\mathbb{E}[\langle DF, -DL^{-1}F \rangle|F] \geq 0$. This proves Theorem 2.1. \square

4.2. *Proof of Proposition 2.3.* We start by observing that the function $x \mapsto \Phi(x) := x|x| = \text{sign}(x)x^2$, $x \in \mathbb{R}$, is such that, for every $a, b \in \mathbb{R}$, $\Phi(b) = \Phi(a) + 2|a|(b - a) + R(a, b)$, where $|R(a, b)| \leq (b - a)^2$. It follows that

$$(\Phi(b) - \Phi(a))^2 \leq 8a^2(b - a)^2 + 2(b - a)^4. \quad (4.5)$$

Since μ is finite,

$$\mathbb{E} \int_{\mathcal{Z}} (D_z\mathbf{1}_{\{F > -\nu\}})^2 \mu(dz) \leq \mathbb{E} \int_{\mathcal{Z}} (\mathbf{1}_{\{F+D_zF > -\nu\}} - \mathbf{1}_{\{F > -\nu\}})^2 \mu(dz) \leq \mu(\mathcal{Z}) < \infty,$$

which implies that $\mathbf{1}_{\{F > -\nu\}} \in \text{dom } D$; see Lemma 3.1 and compare with Remark 2.2 (i). Moreover, our assumptions imply that $DF|DF| = \Phi(DF) \in \text{dom } \delta$. We can now apply the integration by parts formula (3.4), together with the relation

$L^{-1}F = -q^{-1}F$, to deduce that

$$\begin{aligned} q \times A_3(F) &= \mathbb{E} \int_{\mathcal{Z}} (D_z \mathbf{1}_{\{F > -\nu\}}) \Phi(D_z F) \mu(dz) \\ &= \mathbb{E}[\mathbf{1}_{\{F > -\nu\}} \delta(\Phi(DF))] \\ &\leq [\mathbb{E}[\delta(\Phi(DF))^2]]^{1/2}. \end{aligned}$$

Again in view of our assumptions, the Skorohod isometry implied by Privault (2009, Proposition 6.5.4) is verified, and we deduce that

$$\begin{aligned} \mathbb{E}[\delta(\Phi(DF))^2] &\leq \mathbb{E} \int_{\mathcal{Z}} \Phi(D_z F)^2 \mu(dz) + \mathbb{E} \int_{\mathcal{Z}} \int_{\mathcal{Z}} [D_{z_2} \Phi(D_{z_1} F)]^2 \mu(dz_1) \mu(dz_2) \\ &= \mathbb{E} \int_{\mathcal{Z}} (D_z F)^4 \mu(dz) + \mathbb{E} \int_{\mathcal{Z}} \int_{\mathcal{Z}} [D_{z_2} \Phi(D_{z_1} F)]^2 \mu(dz_1) \mu(dz_2). \end{aligned}$$

Since $D_{z_2} \Phi(D_{z_1} F) = \Phi(D_{z_1} F + D_{z_2} D_{z_1} F) - \Phi(D_{z_1} F)$, we can now apply (4.5) with $a = D_{z_1} F$ and $b = D_{z_1} F + D_{z_2} D_{z_1} F$ to infer the upper bound

$$[D_{z_2} \Phi(D_{z_1} F)]^2 \leq 8(D_{z_1} F)^2 (D_{z_2} D_{z_1} F)^2 + 2(D_{z_2} D_{z_1} F)^4,$$

and the conclusion follows immediately. \square

4.3. Proof of Theorem 2.6. Let $F_n = I_q(f_n)$ be as in the statement of Theorem 2.6. Then $\langle DF_n, -DL^{-1}F_n \rangle = \frac{1}{q} \|DI_q(f_n)\|^2$ and $\mathbb{E}[\langle DF_n, -DL^{-1}F_n \rangle | F_n] \geq 0$. Thus, we need to prove that for such F_n the right-hand side of (2.2) converges to zero as $n \rightarrow \infty$. We do this by showing that the three terms $A'_1(F_n)$, $A_3(F_n)$ and $A_4(F_n)$ (see (2.3)) all converge to zero as $n \rightarrow \infty$; the computations performed below will also implicitly provide the upper bound (2.6). It is important to note that our analysis of the terms $A'_1(F_n)$ and $A_4(F_n)$ does not make use of the fact that $\mu_n(\mathcal{Z}) < \infty$. It is convenient to start with the reminder term $A_4(F_n)$.

Lemma 4.1. *Under the conditions of Theorem 2.6, it holds that $A_4(I_q(f_n)) \rightarrow 0$, as $n \rightarrow \infty$.*

Proof: First observe that in our case

$$A_4(I_q(f_n)) = \sqrt{\int_{\mathcal{Z}} \mathbb{E}[|D_z I_q(f_n)|^4] \mu_n(dz)}.$$

We can now use Peccati et al. (2010, formulae (4.17) and (4.18)) to deduce that

$$\begin{aligned} A_4(I_q(f_n)) &\leq q^2 \sum_{r=1}^q \sum_{\ell=0}^{r-1} \mathbf{1}(1 \leq r + \ell \leq 2q - 1) \\ &\quad \times ((r + \ell - 1)!)^{1/2} (q - \ell - 1)! \binom{q-1}{q-1-\ell}^2 \binom{q-1-\ell}{q-r} \|f_n \star_r^\ell f_n\|. \end{aligned} \tag{4.6}$$

Since this estimate does not involve the middle contraction $f_n \star_{q/2}^{q/2} f_n$, the conclusion follows immediately. \square

Now we study the convergence of the sequence $A'_1(F_n)$.

Lemma 4.2. *Under the conditions of Theorem 2.6 we have $A'_1(I_q(f_n)) \rightarrow 0$, as $n \rightarrow \infty$.*

Proof: One must prove that $\mathbb{E}[\|DI_q(f_n)\|^2 - 2qI_q(f_n) - 2q\nu]^2 \rightarrow 0$. Expanding the square and using the fact that $\mathbb{E}[I_q(f_n)] = 0$ we have to show that

$$\begin{aligned} &\mathbb{E}[\|DI_q(f_n)\|^4] - 4q\mathbb{E}[I_q(f_n)\|DI_q(f_n)\|^2] \\ &\quad + 4q^2\mathbb{E}[I_q^2(f_n)] - 4q\nu\mathbb{E}[\|DI_q(f_n)\|^2] + 4q^2\nu^2 \rightarrow 0 \end{aligned} \tag{4.7}$$

as $n \rightarrow \infty$. Firstly, $\mathbb{E}[I_q^2(f_n)] = q!\|f_n\|^2 \rightarrow 2\nu$. The definition of $DI_q(f_n)$ and formula (3.6) imply that

$$\|DI_q(f_n)\|^2 = q q! \|f_n\|^2 + q^2 \sum_{p=1}^{2(q-1)} \int_{\mathcal{Z}} I_p(G_p^{q-1} f_n(z, \cdot)) \mu_n(dz) \tag{4.8}$$

so that $\mathbb{E}[\|DI_q(f_n)\|^2] = q q! \|f_n\|^2$, which asymptotically behaves like $2q\nu$. Using integration by parts (3.4) together with the relation $DF^2 = 2FDF + (DF)^2$ applied to $F = I_q(f)$, we infer that

$$\mathbb{E}[I_q(f_n)\|DI_q(f_n)\|^2] = \frac{q}{2}\mathbb{E}[I_q^3(f_n)] - \frac{1}{2}\mathbb{E} \int_{\mathcal{Z}} D_z I_q^3(f_n) \mu_n(dz).$$

Now, in view of the estimate (4.6), the second summand on the right-hand side of the previous equation converges to zero as $n \rightarrow \infty$, and consequently

$$\mathbb{E}[I_q(f_n)\|DI_q(f_n)\|^2]$$

behaves asymptotically as $\frac{q}{2}\mathbb{E}[I_q^3(f_n)]$. Using (3.5) and the orthogonality of chaoses we obtain

$$\begin{aligned} \mathbb{E}[I_q^3(f_n)] &= \sum_{p=0}^q p! \binom{q}{p}^2 \sum_{\ell=0}^p \binom{p}{\ell} \mathbb{E}[I_{2q-p-\ell}(f_n)I_q(f_n)] \\ &= \sum_{p=q/2}^q p! \binom{q}{p}^2 \binom{p}{q-p} q! \langle f_n \tilde{\star}_p^{q-p} f_n, f_n \rangle, \end{aligned}$$

so that $\mathbb{E}[I_q(f_n)\|DI_q(f_n)\|^2]$ has the same limit as

$$\frac{q}{2} \sum_{p=q/2}^q p! \binom{q}{p}^2 \binom{p}{q-p} q! \langle f_n \tilde{\star}_p^{q-p} f_n, f_n \rangle.$$

Moreover, one can show that

$$\mathbb{E}[\|DI_q(f_n)\|^4] = q^2(q!\|f_n\|^2)^2 + q^4 \sum_{p=1}^{2(q-1)} p! \|\hat{G}_p^q f_n\|^2, \tag{4.9}$$

where $\hat{G}_p^q f_n$ with $p \in \{1, \dots, 2(q-1)\}$ is defined by

$$\hat{G}_p^q f_n = \sum_{t=1}^q \sum_{s=1}^{\min(t, q-1)} \mathbf{1}(2q-t-s=p) (t-1)! \binom{q-1}{t-1}^2 \binom{t-1}{s-1} f_n \tilde{\star}_t^s f_n.$$

Indeed, use (4.8), the orthogonality of the random variables

$$\int_{\mathcal{Z}} I_{p_1}(G_{p_1}^{q-1} f_n(z, \cdot)) \mu_n(dz) \quad \text{and} \quad \int_{\mathcal{Z}} I_{p_2}(G_{p_2}^{q-1} f_n(z, \cdot)) \mu_n(dz)$$

for $1 \leq p_1 \neq p_2 \leq 2(q-1)$ as well as the stochastic Fubini theorem Peccati and Taqqu (2011, Theorem 5.13.1) (which is valid thanks to our technical assumptions

made in Section 3) to conclude that the identity (4.9) is verified; see also the proof of Theorem 4.2 in Peccati et al. (2010). We now exploit the assumption that $\|f_n \star_r^\ell f_n\| \rightarrow 0$ with r and ℓ as in the statement of Theorem 2.6. It implies that

$$\langle f_n \widetilde{\star}_p^{q-p} f_n, f_n \rangle \rightarrow 0 \quad \text{and} \quad \langle f_n \widetilde{\star}_t^s f_n, f_n \widetilde{\star}_{t'}^{s'} f_n \rangle \rightarrow 0 \quad (4.10)$$

as $n \rightarrow \infty$ for all $p \in \{q/2+1, \dots, q\}$ and $t, t' \in \{1, \dots, q\}$, $s \in \{1, \dots, \min(t, q-1)\}$, $s' \in \{1, \dots, \min(t', q-1)\}$ and t, s, t', s' not equal to $q/2$ at the same time. Indeed,

$$|\langle f_n \widetilde{\star}_p^{q-p} f_n, f_n \rangle| \leq \|f_n \widetilde{\star}_p^{q-p} f_n\| \|f_n\| \leq \|f_n \star_p^{q-p} f_n\| \|f_n\| \rightarrow 0$$

for $p \in \{q/2+1, \dots, q\}$ and similarly

$$|\langle f_n \widetilde{\star}_t^s f_n, f_n \widetilde{\star}_{t'}^{s'} f_n \rangle| \leq \|f_n \widetilde{\star}_t^s f_n\| \|f_n \widetilde{\star}_{t'}^{s'} f_n\| \leq \|f_n \star_t^s f_n\| \|f_n \star_{t'}^{s'} f_n\| \rightarrow 0,$$

where t, s, t', s' are as above. Plugging the expressions for $\mathbb{E}[\|DI_q(f_n)\|]$,

$$\mathbb{E}[I_q(f_n)\|DI_q(f_n)\|^2]$$

and $\mathbb{E}[\|DI_q(f_n)\|^2]$ into (4.7) and using the first statement in (4.10) we see immediately that (4.7) has the same limit as

$$8q^2\nu - 2q^2 \left(\frac{q}{2}\right)! \binom{q}{q/2}^2 q! \langle f_n \widetilde{\star}_{q/2}^{q/2} f_n, f_n \rangle + q^4 \sum_{p=1}^{2(q-1)} p! \|\hat{G}_p^q f_n\|^2. \quad (4.11)$$

We notice now that the middle contraction in the sum in (4.11) can only appear in the term $p = q$. Using the definition of $\hat{G}_q^q f_n$ and the second statement in (4.10) we see that $q^4 q! \|\hat{G}_q^q f_n\|^2$ behaves asymptotically like

$$q^4 q! \left(\left(\frac{q}{2} - 1\right)!\right)^2 \binom{q-1}{q/2-1}^4 \|f_n \widetilde{\star}_{q/2}^{q/2} f_n\|^2.$$

Consequently, (4.11) has the same limit as

$$\begin{aligned} & (8q^2\nu - 4q^2 q! \|f_n\|^2) + 4q^2 q! \|f_n\|^2 - 2q^2 \left(\frac{q}{2}\right)! \binom{q}{q/2}^2 q! \langle f_n \widetilde{\star}_{q/2}^{q/2} f_n, f_n \rangle \\ & + q^4 q! \left(\left(\frac{q}{2} - 1\right)!\right)^2 \binom{q-1}{q/2-1}^4 \|f_n \widetilde{\star}_{q/2}^{q/2} f_n\|^2 \rightarrow 0, \end{aligned}$$

as $n \rightarrow \infty$, where we have used the fact that

$$\|f_n \widetilde{\star}_{q/2}^{q/2} f_n\|^2 \rightarrow \frac{2}{q!} c_q^2 \nu,$$

and

$$\langle f_n \widetilde{\star}_{q/2}^{q/2} f_n, f_n \rangle \rightarrow \frac{2}{q!} c_q \nu.$$

This proves the claim. \square

We eventually deal with the convergence of the sequence

$$A_3(I_q(f_n)) = \frac{1}{q} \int_{\mathcal{Z}} \mathbb{E}[(D_z \mathbf{1}_{\{I_q(f_n) > -\nu\}}) D_z I_q(f_n) | D_z I_q(f_n)] \mu_n(dz), \quad n \geq 1.$$

Lemma 4.3. *Under the conditions of Theorem 2.6 we have that $A_3(I_q(f_n)) \rightarrow 0$, as $n \rightarrow \infty$.*

Proof: In view of the assumptions, we can directly apply Proposition 2.3. It follows that our claim is proved once we show that the three terms on the right-hand side of (2.4) (with $F = F_n = I_q(f_n)$ and $\mu = \mu_n$) converge to zero as $n \rightarrow \infty$. Since the first term equals $A_4(F_n)$, by virtue of the previous Lemma 4.1, we only have to prove the convergence of the remaining two summands. Our starting point is the following representation of the quantity $(D_{z_2}D_{z_1}F_n)^2 = q^2(q-1)^2 I_{q-2}^2(f_n(z_1, z_2, \cdot))$, which is obtained by means of the product formula (3.5). Indeed,

$$\begin{aligned} (D_{z_2}D_{z_1}F_n)^2 &= q^2(q-1)^2 \sum_{r=0}^{q-2} \sum_{\ell=0}^r r! \binom{q-2}{r}^2 \binom{r}{\ell} \\ &\quad \times I_{2(q-2)-r-\ell}(f_n(z_1, z_2, \cdot)) \star_r^\ell f_n(z_1, z_2, \cdot) \\ &= q^2(q-1)^2 I_{q-2}^2(f_n(z_1, z_2, \cdot)). \end{aligned}$$

Combining this representation with an iterated application of the triangle inequality, as well as of the isometric properties of multiple integrals, one deduces that the quantity

$$\sqrt{\mathbb{E} \int_{\mathcal{Z}} \int_{\mathcal{Z}} (D_{z_2}D_{z_1}F_n)^4 \mu_n(dz_1)\mu_n(dz_2)}$$

is bounded by a linear combination (with coefficients not depending on n) of quantities of the type

$$\sqrt{\int_{\mathcal{Z}} \int_{\mathcal{Z}} \|f_n(z_1, z_2, \cdot) \star_r^\ell f_n(z_1, z_2, \cdot)\|^2 \mu_n(dz_1)\mu_n(dz_2)} = \|f_n \star_{q-\ell}^{q-2-r} f_n\| \rightarrow 0,$$

where the equality follows from a standard application of Fubini’s theorem, and the convergence to zero is a consequence of the fact that $a := q - \ell \in \{2, \dots, q\}$ and $b := q - 2 - r \in \{0, \dots, a - 2\}$, as well as of the elementary identity $\|f_n \star_a^0 f_n\| = \|f_n \star_q^{q-a} f_n\|$ ($2 \leq a \leq q$). To deal with the remaining middle term, we use Fubini’s theorem and the Cauchy-Schwarz inequality to deduce the estimate

$$\sqrt{\mathbb{E} \int_{\mathcal{Z}} \int_{\mathcal{Z}} (D_{z_2}D_{z_1}F_n)^2 (D_{z_1}F_n)^2 \mu_n(dz_1)\mu_n(dz_2)} \leq A_4(F_n)^{1/2} \times C_n^{1/4},$$

with

$$C_n := \mathbb{E} \int_{\mathcal{Z}} \left(\int_{\mathcal{Z}} (D_{z_2}D_{z_1}F_n)^2 \mu_n(dz_2) \right)^2 \mu_n(dz_1).$$

Using again the explicit representation of $(D_{z_2}D_{z_1}F_n)^2$ and applying several times Fubini’s theorem, one sees that C_n is indeed equal to a linear combination (with coefficients not depending on n) of objects of the type

$$\|f_n \star_a^b f_n\|^2, \quad \text{with } a = 2, \dots, q \text{ and } b = 0, \dots, a - 2.$$

The conclusion follows immediately since our estimates do not involve the middle contraction. \square

4.4. *Proof of Proposition 2.9.* The product formula (3.5) shows that

$$\begin{aligned} I_2^2(f_n) &= I_4(f_n \tilde{\star}_0^0 f_n) + 4I_3(f_n \tilde{\star}_1^0 f_n) + I_2(4f_n \star_1^1 f_n + 2f_n^2) \\ &\quad + 4I_1(f_n \star_2^1 f_n) + 2\|f_n\|^2. \end{aligned} \tag{4.12}$$

Using the relation

$$4!\|f_n \tilde{\star}_0 f_n\|^2 = 2(2\|f_n\|^2)^2 + 16\|f_n \star_1 f_n\|^2 \quad (4.13)$$

(see e.g. [Nourdin and Peccati \(2012, formula \(5.2.12\)\)](#)), exploiting the orthogonality of multiple integrals of distinct orders and using the fact that $\|f_n^2\| \rightarrow 0$ by assumption, we infer that $\mathbb{E}[I_2^4(f_n)] - 12\mathbb{E}[I_2^3(f_n)]$ has the same limit as

$$\begin{aligned} & 16 \times 3!\|f_n \tilde{\star}_1 f_n\|^2 + 16\|f_n \star_2 f_n\|^2 + 48\|f_n \star_1 f_n\|^2 \\ & \quad - 96\langle f_n \star_1 f_n, f_n \rangle + 3(2\|f_n\|^2)^2 \\ & = 16 \times 3!\|f_n \tilde{\star}_1 f_n\|^2 + 16\|f_n \star_2 f_n\|^2 \\ & \quad + 48\|f_n \star_1 f_n - f_n\|^2 - 48\|f_n\|^2 + 3(2\|f_n\|^2)^2. \end{aligned}$$

The conclusion follows by observing that $\|f_n\|^2 \rightarrow \nu$ by assumption, and then by applying [Theorem 2.6](#). \square

4.5. Proof of [Theorem 2.13](#). Proof of Part A. According to [Lemma 2.11](#), since each \tilde{F}'_n is completely degenerate, one has that $\tilde{F}'_n = I_2(f_n)$, where $f_n = h_n/\sigma(n)$, and the double integral is performed with respect to the compensated Poisson measure $\hat{\eta}_n = \eta_n - \mu_n$. It follows that the estimate [\(2.12\)](#) is a direct consequence of [Peccati et al. \(2010, Theorem 4.2\)](#). Using formulae [\(4.12\)](#) and [\(4.13\)](#), we deduce that

$$\begin{aligned} \mathbb{E}[I_2^4(f_n)] & = 16 \times 3!\|f_n \tilde{\star}_1 f_n\|^2 + 16\|f_n \star_2 f_n\|^2 + 16\|f_n \star_1 f_n\|^2 \\ & \quad + 2\|4f_n \star_1 f_n + 2f_n^2\|^2 + 3(2\|f_n\|^2)^2, \end{aligned}$$

where the norms and contractions are of course taken with respect to the measure μ_n . Since $3(2\|f_n\|^2)^2$ converges to 3 by assumption, we deduce that, if [\(2.11\)](#) is verified, then the right-hand side of [\(2.12\)](#) converges to zero, and therefore \tilde{F}'_n converges in distribution to N . To conclude, observe that the estimates contained in [Dynkin and Mandelbaum \(1983, pp. 744-745\)](#) yield that $\mathbb{E}[(\tilde{F}'_n - \tilde{F}_n)^2] = O(n^{-1/2})$ as $n \rightarrow \infty$, so that the estimate [\(2.13\)](#) follows from the elementary inequality

$$d_W(\tilde{F}_n, N) \leq d_W(\tilde{F}'_n, N) + [\mathbb{E}(\tilde{F}'_n - \tilde{F}_n)^2]^{1/2}.$$

Proof of Part B. Again in view of [Lemma 2.11](#) and of the complete degeneracy of each F'_n , we deduce that $F'_n = I_2(h_n)$, where the double integral is again with respect to the compensated Poisson measure corresponding to η_n . The estimate [\(2.15\)](#) is therefore a consequence of [Theorem 2.6](#), and the fact that the distribution of \tilde{F}'_n converges to $\bar{\Gamma}_\nu$ is a direct consequence of [Proposition 2.9](#) in the case $h_n = f_n$. The conclusion follows once again from the fact that $\mathbb{E}[(F'_n - F_n)^2] = O(n^{-1/2})$ as $n \rightarrow \infty$, in such a way that [\(2.16\)](#) follows from the triangle inequality

$$d_3(\tilde{F}_n, \bar{\Gamma}_\nu) \leq d_3(\tilde{F}'_n, \bar{\Gamma}_\nu) + [\mathbb{E}(\tilde{F}'_n - \tilde{F}_n)^2]^{1/2}.$$

This completes the proof. \square

4.6. Proof of [Theorem 2.17](#). According to [Lachièze-Rey and Peccati \(2013b, Theorem 7.3\)](#), one has that

$$\tilde{F}'_n = I_2(h_n) + R_n,$$

where $h_n = h_2/n = n^{-1} \sum_{i=1}^\nu e_i \otimes e_i$, the double integral is realized with respect to the compensated Poisson measure $\hat{\eta}_n = \eta_n - n\mu$, and R_n is a residual sequence

of random variables such that

$$\mathbb{E}[R_n^2] = O(1/n), \quad \text{as } n \rightarrow \infty.$$

It is immediate to verify that: (a) $(\int_{\mathcal{Z}} h_n^4 d\mu_n^2)^{1/4} = O(1/\sqrt{n})$ as $n \rightarrow \infty$, (b) $h_n \star_1^1 h_n = h_n$ (where the contraction is realized with respect to μ_n), (c) $\|h_n \star_2^1 h_n\| = O(n^{-1/2})$ as $n \rightarrow \infty$ (since $h_n \star_2^1 h_n(x) = n^{-1} \sum_{i=1}^{\nu} e_i(x)^2$). The estimates are therefore a consequence of Theorem 2.13-(B), as well as of the estimates $\mathbb{E}[(F'_n - F_n)^2] = O(n^{-1/2})$ as $n \rightarrow \infty$ and

$$d_3(\tilde{F}_n, \bar{\Gamma}_\nu) \leq d_3(\tilde{F}'_n, \bar{\Gamma}_\nu) + [\mathbb{E}(\tilde{F}'_n - \tilde{F}_n)^2]^{1/2}.$$

This completes the proof. □

4.7. *Proof of Theorem 2.20.* We start with some general preliminaries which will be specialized below.

Let $F_n^{(1)}, \dots, F_n^{(d)}$ be centered square-integrable functionals of the Poisson measure η in the domain of the derivative operator D . For $i \in \{1, \dots, d\}$ let us define

$$\begin{aligned} \alpha_n^{(i)} &:= \mathbb{E}|2(F_n^{(i)} + \nu_i)_+ - \langle DF_n^{(i)}, -DL^{-1}F_n^{(i)} \rangle| \\ &+ \mathbb{E} \int_{\mathcal{Z}} |D_z F_n^{(i)}|^2 |D_z L^{-1}F_n^{(i)}| \mu_n(dz) \\ &+ \mathbb{E} \int_{\mathcal{Z}} (D_z \mathbf{1}_{\{F_n^{(i)} > -\nu_i\}})(D_z F_n^{(i)}) |D_z L^{-1}F_n^{(i)}| \mu_n(dz), \end{aligned} \tag{4.14}$$

and for $i \neq j \in \{1, \dots, d\}$ put

$$\begin{aligned} \beta_n^{(i,j)} &:= \mathbb{E}|\langle DF_n^{(i)}, DL^{-1}F_n^{(j)} \rangle|, \\ \gamma_n^{(i,j)} &:= \mathbb{E} \int_{\mathcal{Z}} |D_z F_n^{(i)}|^2 |D_z L^{-1}F_n^{(j)}| \mu_n(dz). \end{aligned} \tag{4.15}$$

We estimate the distance between (the law of) $\mathbf{F}_n := (F_n^{(1)}, \dots, F_n^{(d)})$ and (that of) $\mathbf{\Gamma} := (G_1, \dots, G_d)$ by $d(\mathbf{F}_n, \mathbf{\Gamma}) = \sup |\mathbb{E}\phi(\mathbf{F}_n) - \mathbb{E}\phi(\mathbf{\Gamma})|$, where the supremum runs over all functions $\phi : \mathbb{R}^d \rightarrow \mathbb{R}$ whose partial derivatives up to order 3 are bounded, continuous and satisfy $\|\cdot\|_\infty \leq 1$. We notice that if $d(\mathbf{F}_n, \mathbf{\Gamma}) \rightarrow 0$ then $\mathbf{F}_n \xrightarrow{d} \mathbf{\Gamma}$ as $n \rightarrow \infty$.

Lemma 4.4. *There exist constants K_1 and K_2 such that*

$$d(\mathbf{F}_n, \mathbf{\Gamma}) \leq K_1 \sum_{i=1}^d \alpha_n^{(i)} + K_2 \sum_{\substack{i,j=1 \\ i \neq j}}^d (\beta_n^{(i,j)} + \gamma_n^{(i,j)}).$$

Proof: The technique adopted here is similar to the one used in the proof of the main result of Bourguin and Peccati (2012). To keep the argument more transparent and the formulas simpler we restrict ourselves to the case $d = 2$, the general case can be dealt with similarly. So, $\mathbf{F}_n = (F_n^{(1)}, F_n^{(2)})$ and $\mathbf{\Gamma} = (G_1, G_2)$ and we have to show that

$$d((F_n^{(1)}, F_n^{(2)}), (G_1, G_2)) \leq K_1(\alpha_n^{(1)} + \alpha_n^{(2)}) + K_2(\beta_n^{(1,2)} + \beta_n^{(2,1)} + \gamma_n^{(1,2)} + \gamma_n^{(2,1)}). \tag{4.16}$$

To accomplish this task, we shall provide uniform estimates on $|\mathbb{E}\phi(F_n^{(1)}, F_n^{(2)}) - \mathbb{E}\phi(G_1, G_2)|$. First write

$$\begin{aligned} |\mathbb{E}\phi(F_n^{(1)}, F_n^{(2)}) - \mathbb{E}\phi(G_1, G_2)| &\leq |\mathbb{E}[\phi(F_n^{(1)}, F_n^{(2)})] - \mathbb{E}[\phi(G_1, F_n^{(2)})]| \\ &\quad + |\mathbb{E}[\phi(G_1, F_n^{(2)})] - \mathbb{E}[\phi(G_1, G_2)]| =: |T_1| + |T_2|. \end{aligned}$$

We first deal with T_2 . Conditioning on G_1 , we are in a one-dimensional situation and can proceed as in the proof of Theorem 2.1. This shows that T_2 contributes the term $\alpha_n^{(2)}$ to the bound (4.16). We now consider the term T_1 and write \mathcal{L}_U for the law of a random object U . Rewriting yields

$$T_1 = \int \left(\phi(x, y) - \int \phi(g, y) \mathcal{L}_{G_1}(dg) \right) \mathcal{L}_{(F_n^{(1)}, F_n^{(2)})}(d(x, y)).$$

For fixed y we consider the term in brackets as the left-hand side of a Stein-equation for the $\bar{\Gamma}_{\nu_1}$ -distribution so that

$$\begin{aligned} &\int \left(\phi(x, y) - \int \phi(g, y) \mathcal{L}_{G_1}(dg) \right) \mathcal{L}_{(F_n^{(1)}, F_n^{(2)})}(d(x, y)) \\ &= \int 2(x + \nu_1)_+ h'_y(x) - x h_y(x) \mathcal{L}_{(F_n^{(1)}, F_n^{(2)})}(d(x, y)), \end{aligned} \quad (4.17)$$

where, for fixed y , $h_y(x)$ is the solution of the Stein-equation associated with the test function $x \mapsto \phi(x, y)$. We now consider the bivariate function $\hat{h}(x, y) := h_y(x)$. Using the smoothness assumptions on ϕ together with the explicit representation

$$\hat{h}(x, y) = \begin{cases} -\frac{1}{x} (\phi(x, y) - \mathbb{E}[\phi(G(\nu_1), y)]) & : x \leq -\nu_1 \\ \frac{1}{2(x + \nu_1)_+ g_{\nu_1}(x)} \int_{-\nu_1}^x (\phi(z, y) - \mathbb{E}[\phi(G(\nu_1), y)]) g_{\nu_1}(z) dz & : x > -\nu_1, \end{cases}$$

(recall the discussion preceding the proof of Theorem 2.1 and notice that $g_{\nu_1}(\cdot)$ stands for the density of the law $\bar{\Gamma}_{\nu_1}$) we deduce the following facts: (i) the mapping $x \mapsto \hat{h}(x, y)$ (for fixed y) is twice differentiable on $\mathbb{R} \setminus \{-\nu\}$ (and it also admits right and left first derivatives at $x = -\nu$), and (ii) the mapping $y \mapsto \hat{h}(x, y)$ (for fixed x) is twice differentiable on \mathbb{R} . All the involved derivatives are bounded by a finite constant only depending on ν_1 . Note that, in order to establish the estimates on $y \mapsto \hat{h}(x, y)$, one has to take derivatives under the integral and expectation signs, which is allowed thanks to the assumptions on ϕ .

After these technical considerations we observe that (4.17) may be expressed in terms of \hat{h} as

$$\begin{aligned} &\int 2(x + \nu_1)_+ \partial_1 \hat{h}(x, y) - x \hat{h}(x, y) \mathcal{L}_{(F_n^{(1)}, F_n^{(2)})}(d(x, y)) \\ &= \mathbb{E}[2(F_n^{(1)} + \nu_1)_+ \partial_1 \hat{h}(F_n^{(1)}, F_n^{(2)}) - F_1 \hat{h}(F_n^{(1)}, F_n^{(2)})] \\ &= \mathbb{E}[2(F_n^{(1)} + \nu_1)_+ \partial_1 \hat{g}(F_n^{(1)}, F_n^{(2)}) - \langle D\hat{h}(F_n^{(1)}, F_n^{(2)}), -DL^{-1}F_n^{(1)} \rangle], \end{aligned} \quad (4.18)$$

where ∂_1 stands for the partial derivative with respect to the first coordinate and where we have applied the integration by parts formula (3.4) of Malliavin calculus in exactly the same way as in the proof of Theorem 2.1. Using the notation $F_{n,z}^{(i)}(\eta) =$

$F_n^{(i)}(\eta + \delta_z) - F_n^{(i)}(\eta)$ for $i \in \{1, 2\}$ and $z \in \mathcal{Z}$, we may write

$$\begin{aligned} D_z \hat{h}(F_1, F_2) &= \hat{h}(F_{n,z}^{(1)}, F_{n,z}^{(2)}) - \hat{h}(F_n^{(1)}, F_n^{(2)}) \\ &= (\hat{h}(F_{n,z}^{(1)}, F_{n,z}^{(2)}) - \hat{h}(F_n^{(1)}, F_n^{(2)})) + (\hat{h}(F_{n,z}^{(1)}, F_n^{(2)}) - \hat{h}(F_n^{(1)}, F_n^{(2)})) \\ &=: S_1 + S_2 \end{aligned}$$

Thanks to the properties of \hat{h} described above, we find that

$$S_1 = \partial_2 \hat{h}(F_{n,z}^{(1)}, F_n^{(2)}) D_z F_n^{(2)} + R^{(1)}(D_z F_n^{(2)})$$

and

$$S_2 = \partial_1 \hat{h}(F_n^{(1)}, F_{n,z}^{(2)}) D_z F_n^{(1)} + R^{(2)}(D_z F_n^{(1)}),$$

where $R^{(1)}$ and $R^{(2)}$ are such that

$$|R^{(1)}(D_z F_n^{(2)})| \leq K_1^{(1)} |D_z F_n^{(2)}|^2$$

and

$$|R^{(2)}(D_z F_n^{(1)})| \leq K_1^{(2)} |D_z F_n^{(1)}|^2 + K_2^{(2)} (D_z \mathbf{1}_{\{F_n^{(1)} > -\nu_1\}})(D_z F_n^{(1)}),$$

where ∂_{11} and ∂_{22} , respectively, denote the second derivative with respect to the first and second coordinate and where $K_1^{(1)}, K_1^{(2)}, K_2^{(2)}$ are finite constants. Combining this with (4.18) and taking the supremum over all ϕ , we obtain the contributions $\alpha_n^{(1)}, \beta_n^{(2,1)}$ and $\gamma_n^{(2,1)}$ in (4.16). Inverting the role of $F_n^{(1)}$ and $F_n^{(2)}$ in the previous discussion gives the bound (4.16), with constants K_1 and K_2 only depending on (ν_1, ν_2) . \square

Proof of Theorem 2.20: Let us define the random vector

$$\mathbf{I}_n := (I_{q_1}(f_n^{(i)}), \dots, I_{q_d}(f_n^{(d)})).$$

We shall prove that $d(\mathbf{I}_n, \mathbf{\Gamma}) \rightarrow 0$ as $n \rightarrow \infty$. Lemma 4.4 implies that for this it is sufficient to check that $\alpha_n^{(i)} \rightarrow 0, \beta_n^{(i,j)} \rightarrow 0$ and that $\gamma_n^{(i,j)} \rightarrow 0$ as $n \rightarrow \infty$ for any combination of i and j . Under the assumptions in the statement, writing $F_n^{(i)} = I_{q_i}(f_n^{(i)})$ one has the following three facts for every $i = 1, \dots, d$: (a) $\alpha_n^{(i)} \rightarrow 0$, as $n \rightarrow \infty$, (b) as $n \rightarrow \infty$,

$$\mathbb{E} \int_{\mathcal{Z}} (D_z F_n^{(i)})^4 \mu_n(dz) \rightarrow 0,$$

and (c) the sequence

$$\mathbb{E} \int_{\mathcal{Z}} (D_z F_n^{(i)})^2 \mu_n(dz) = q_i^2 \mathbb{E} \int_{\mathcal{Z}} (DL^{-1} F_n^{(i)})^2 \mu_n(dz), \quad n \geq 1,$$

is bounded. An application of the Cauchy-Schwarz inequality yields therefore that $\gamma_n^{(i,j)} \rightarrow 0$ for any allowed choice of i and j . To check the fact that $\beta_n^{(i,j)} \rightarrow 0$, we apply once more the Cauchy-Schwarz inequality to obtain

$$\beta_n^{(i,j)} \leq q_i^2 \left(\mathbb{E} \left(\int_{\mathcal{Z}} I_{q_i-1}(f_n^{(i)}(z, \cdot)) I_{q_j-1}(f_n^{(j)}(z, \cdot)) \mu_n(dz) \right)^2 \right)^{1/2}.$$

We use now the general product formula (3.5) for multiple integrals to express

$$I_{q_i-1}(f_n^{(i)}(z, \cdot)) I_{q_j-1}(f_n^{(j)}(z, \cdot))$$

as a sum of multiple integrals and the stochastic version of Fubini's theorem allowing us to exchange deterministic with stochastic integration; see Peccati and Taqqu

(2011, Theorem 5.13.1). By assumption, $q_i < q_j$. Using the triangle inequality several times yields

$$\begin{aligned} & \left(\mathbb{E} \left(\int_{\mathcal{Z}} I_{q_i-1}(f_n^{(i)}(z, \cdot)) I_{q_j-1}(f_n^{(j)}(z, \cdot)) \mu_n(dz) \right)^2 \right)^{1/2} \\ & \leq \sum_{r=1}^{q_i} \sum_{\ell=1}^r K(r, \ell, q_i, q_j)^{1/2} \|f_n^{(i)} \star_r^\ell f_n^{(j)}\|, \end{aligned}$$

with the constant $K(r, \ell, q_i, q_j)$ given by

$$K(r, \ell, q_i, q_j) = (r-1)! \binom{q_i-1}{r-1} \binom{q_j-1}{r-1} \binom{r-1}{\ell-1} (q_i + q_j - r - \ell)!.$$

The proof is completed by observing that (see Peccati and Zheng (2010, Lemma 2.9))

$$\|f_n^{(i)} \star_r^\ell f_n^{(j)}\| \leq \|f_n^{(i)} \star_r^\ell f_n^{(i)}\| \|f_n^{(j)} \star_r^\ell f_n^{(j)}\| \rightarrow 0$$

for all choices of i, j , because of the assumptions in the theorem and the fact that $2q_i \neq q_j$ for $i \neq j$. \square

4.8. *Proof of Theorem 2.21.* We start again with some preliminaries. Let

$$F_n^{(1)}, \dots, F_n^{(d_1+d_2)}$$

be square integrable functionals of the Poisson measure η . For $i \in \{d_1+1, \dots, d_1+d_2\}$ let us define

$$\delta_n^{(i)} := \mathbb{E}|1 - \langle DF_n^{(i)}, -DL^{-1}F_n^{(i)} \rangle| + \mathbb{E} \int_{\mathcal{Z}} |D_z F_n^{(i)}|^2 |D_z L^{-1}F_n^{(i)}| \mu_n(dz)$$

and for $i \in \{1, \dots, d_1\}$ let $\alpha_n^{(i)}$ be as in (4.14) and for $i, j \in \{1, \dots, d_1+d_2\}$ let $\beta_n^{(i)}$ and $\gamma_n^{(i)}$ be as in (4.15). We will estimate the distance between (the law of) $\mathbf{F}_n := (F_n^{(1)}, \dots, F_n^{(d_1+d_2)})$ and (that of) the hybrid vector

$$\mathbf{H} := (G_1, \dots, G_{d_1}, N_{d_1+1}, \dots, N_{d_2})$$

by the hybrid distance $d_h(\mathbf{F}_n, \mathbf{H}) = \sup |\mathbb{E}\phi(\mathbf{F}_n) - \mathbb{E}\phi(\mathbf{H})|$, where the supremum runs over all functions

$\phi : \mathbb{R}^{d_1+d_2} \rightarrow \mathbb{R}$ whose partial derivatives up to order 3 are bounded, continuous and satisfy $\|\cdot\|_\infty \leq 1$.

Lemma 4.5. *There exist constants K_1, K_2 and K_3 such that*

$$d_h(\mathbf{F}_n, \mathbf{H}) \leq K_1 \sum_{i=1}^{d_1} \alpha_n^{(i)} + K_2 \sum_{i=d_1+1}^{d_1+d_2} \delta_n^{(i)} + K_3 \sum_{\substack{i,j=1 \\ i \neq j}}^{d_1+d_2} (\beta_n^{(i,j)} + \gamma_n^{(i,j)}).$$

Proof: This follows along the same lines of argumentation as the proof of Lemma 4.4. For this reason the details are omitted. \square

Proof of Theorem 2.21: We first use Lemma 4.2 to see that because of (2.17), $\alpha_n^{(i)} \rightarrow 0$ for any $i \in \{1, \dots, d_1\}$. Next, we apply Peccati et al. (2010, Theorem 5.1) to infer that under (2.18), $\delta_n^{(i)} \rightarrow 0$ as $n \rightarrow \infty$ for any $i \in \{d_1+1, \dots, d_1+d_2\}$. The remaining discussion of $\beta_n^{(i,j)}$ and $\gamma_n^{(i,j)}$ is very similar to the multivariate pure

Gamma case so that $\beta_n^{(i,j)} \rightarrow 0$ and $\gamma_n^{(i,j)} \rightarrow 0$ for all $i \neq j \in \{1, \dots, d_1 + d_2\}$. In view of Lemma 4.5, this completes the proof. \square

4.9. *Proof of Theorem 2.23.* We consider a measurable bounded test function $\phi : \mathbb{R} \times \mathbb{Z}_+ \rightarrow \mathbb{R}$ such that ϕ has uniformly bounded derivatives up to the order three in the first variable. By a slight variation of the arguments leading to the proof of Bourguin and Peccati (2012, Theorem 2.1) one has that there exists a universal constant $K > 0$ (independent of n) such that

$$|\mathbb{E}[\phi(I_q(f_n), H_n)] - \mathbb{E}[\phi(G, P)]| \leq K (A_n + B_n + C_n + D_n),$$

where (similar to α_n etc. above)

$$\begin{aligned} A_n &:= \mathbb{E} |2(F_n + \nu)_+ - \langle DF_n, -DL^{-1}F_n \rangle| \\ &\quad + \int_{\mathcal{Z}} \mathbb{E}[|D_z F_n|^2 |D_z L^{-1}F_n|] \mu_n(dz), \\ B_n &:= |\mathbb{E}[H_n] - \lambda| + \mathbb{E} |\lambda - \langle DH_n, -DL^{-1}H_n \rangle| \\ &\quad + \int_{\mathcal{Z}} \mathbb{E}[|D_z H_n(D_z H_n - 1)D_z L^{-1}H_n|] \mu_n(dz), \\ C_n &:= \mathbb{E}[|\langle DH_n, |DI_q(f_n)| \rangle|] \end{aligned}$$

and

$$D_n := \mathbb{E} \int_{\mathcal{Z}} (D_z \mathbf{1}_{\{I_q(f_n) > -\nu\}})(D_z I_q(f_n)) |D_z L^{-1}H_n| \mu_n(dz)$$

In view of Theorem 2.6 (as well as of the estimates leading to its proof), the assumptions in the statement imply that $A_n + B_n + D_n \rightarrow 0$, and, moreover, that

$$\mathbb{E} \int_{\mathcal{Z}} (D_z I_q(f_n))^4 \mu_n(dz) \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

The conclusion is obtained by observing that, by virtue of Hölder's inequality, and since DH_n takes values in \mathbb{Z} ,

$$\begin{aligned} C_n &\leq \left(\mathbb{E} \int_{\mathcal{Z}} (D_z I_q(f_n))^4 \mu_n(dz) \right)^{1/4} \times \left(\mathbb{E} \int_{\mathcal{Z}} (D_z H_n)^{4/3} \mu_n(dz) \right)^{3/4} \\ &\leq \left(\mathbb{E} \int_{\mathcal{Z}} (D_z I_q(f_n))^4 \mu_n(dz) \right)^{1/4} \times \left(\mathbb{E} \int_{\mathcal{Z}} (D_z H_n)^2 \mu_n(dz) \right)^{3/4} \rightarrow 0, \end{aligned}$$

where we have implicitly used assumption (2.19). \square

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