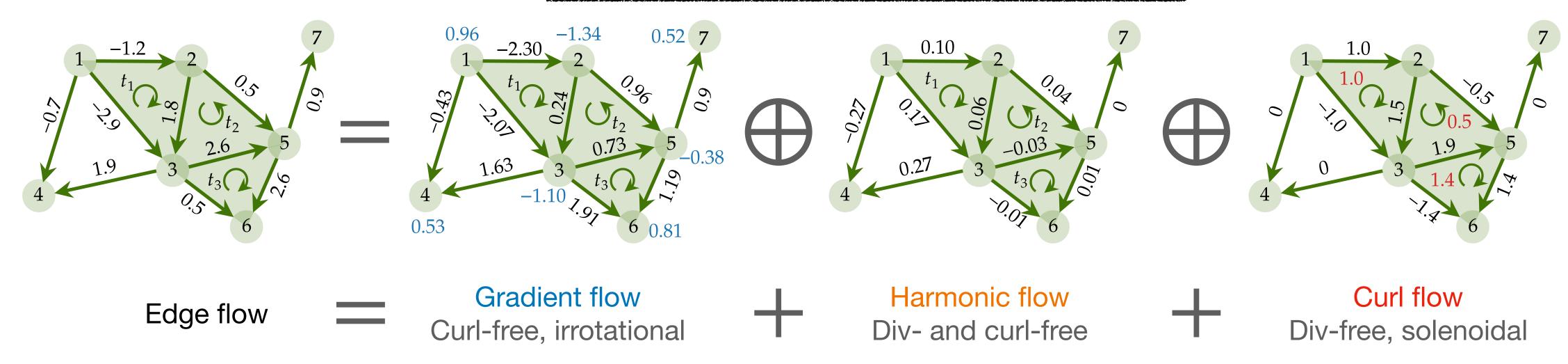
Hodge decomposition

Lovász et al. 2004; Lim et al. 2020

$$\mathbb{R}^{N_1} = \operatorname{im}(\mathbf{B}_1^{\mathsf{T}}) \oplus \ker(\mathbf{L}_1) \oplus \operatorname{im}(\mathbf{B}_2)$$
$$\mathbf{f}_1 = \mathbf{f}_G + \mathbf{f}_H + \mathbf{f}_C$$





Hodge-compositional Edge GP

$$\mathbf{f}_{G} \sim \mathrm{GP}(\mathbf{0}, \mathbf{K}_{G})$$
 $\mathbf{f}_{H} \sim \mathrm{GP}(\mathbf{0}, \mathbf{K}_{H})$
 $\mathbf{f}_{C} \sim \mathrm{GP}(\mathbf{0}, \mathbf{K}_{C})$

Hodge-compositional Edge GPs

Curl-free, div-free GPs

$$\mathbf{f}_{G} \sim \mathrm{GP}(\mathbf{0}, \mathbf{K}_{G})$$
 $\mathbf{f}_{H} \sim \mathrm{GP}(\mathbf{0}, \mathbf{K}_{H})$
 $\mathbf{f}_{C} \sim \mathrm{GP}(\mathbf{0}, \mathbf{K}_{C})$

• Gradient kernel $\mathbf{K}_G = \mathbf{U}_G \Psi_G(\Lambda_G) \mathbf{U}_G^{\mathsf{T}}$; Curl kernel $\mathbf{K}_C = \mathbf{U}_C \Psi_C(\Lambda_C) \mathbf{U}_C^{\mathsf{T}}$

• Matérn family:
$$\Psi_{\square}(\Lambda_{\square}) = \sigma_{\square}^2 \left(\frac{2\nu_{\square}}{\kappa_{\square}^2}\mathbf{I} + \Lambda_{\square}\right)^{-\nu_{\square}}, \quad \square = H, G, C$$

Also as solutions of SDEs, e.g.,

 $\Phi_C(\mathbf{L}_{1,\mathrm{u}})\,\mathbf{f}_1=\mathbf{w}_C$, with curl noise $\mathbf{w}_C\sim N(0,\sigma_C^2\mathbf{U}_C\mathbf{U}_C^{\mathsf{T}})$ and

$$\Phi(\mathbf{L}_{1,u}) = \left(\frac{2\nu_C}{\kappa_C^2}\mathbf{I} + \mathbf{L}_{1,u}\right)^{\frac{\nu_C}{2}} \text{or } \Phi(\mathbf{L}_{1,u}) = e^{\frac{\kappa_C^2}{4}\mathbf{L}_{1,u}}$$