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Non-gaussian geometrical measures in redshift space

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Summary

Density in redshift space

$$\rho_s(\mathbf{X}, \nu) = \int dz \,\rho(\mathbf{x}) \,\frac{1}{\sqrt{2\pi\beta_T(\mathbf{x})}} \exp\left[-\frac{(\nu - \nu_c(\mathbf{x}) - u(\mathbf{x}))^2}{2\beta_T(\mathbf{x})}\right]$$

Cold flow: $\beta_T \to 0$ (effectivly never achieved due to finite redshift rezolution) $\rho_s(\mathbf{X}, v) = \int dz \ \rho(\mathbf{x}) \ \delta_D \left(v - v_c(\mathbf{x}) - u(\mathbf{x}) \right)$

ISM/turbulence studies

No detailed predictions from underlying theory: mode content or relation between velocities and density are not known apriori.

Lesson: slice and dice PPV cube, synthetically varying some control parameters to disentangle effects. Example: measuremets in velocity channels of variable width.

Cosmology

Good predictive understanding of theory that lead to PPZ. Natural approach: try to fit full predictions of power and higher order spectra.

Still valuable - measure some integrals of spectra that take us straight to cosmological info. Having control parameters to vary in your analysis helps.

Geometrical measures for random fields



- Properties of $\nu = const$ isocontours Minkowski functionals (genus/Euler characteristics $\chi(\nu)$, length of isocontours, pencil beam isocontour crossing statistics, $N_D(\nu)$)
- Statistics of extrema peaks, minima, saddles
- Skeleton of the structure, its statistics (length, curvature)
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Geometrical measures for random fields

Starting point

Let us think about such properties of a random field ρ as Euler characteristic (genus), density of maxima, length of skeleton. Their computation reguire knowledge of the joint distribution

 $\mathcal{P}(\rho, \rho_i, \rho_{ij}, \ldots)$

of the field ρ and its first ρ_i , second ρ_{ij} (Hessian matrix) and perhaps higher derivatives, for instance

$$n_{max}(v) = \int_{0 \ge \lambda_1 \ge \lambda_2 \ge \dots} \mathscr{P}(\rho = v, \rho_i, \rho_{ij}) \delta(\rho_i) |\rho_{ij}| d\rho_{ij}$$

Usual approach is to deal with it in the Hessian eigenvalue space, since that's where the boundary conditions are the simplest.

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Non-Gaussian expansion for geometrical statistics

• To treat non-Gaussianities the idea is to expand $\mathscr{P}(\rho, \rho_i, \rho_{ij})$ into orthogonal polinomials around the Gaussian approximation like

$$P(x) = G(x) \left(1 + \sum_{n} \langle x^{n} \rangle_{c} H_{n}(x) \right)$$

- The trick to avoid difficulties is an appropriate choice of variables:
 - that are invariant wrt symmetries of the problem (isotropy)
 - that are polynomial in the field quanities (λ's are no good)
 - · that simplify the Gaussian limit, being as uncorrelated as possible
- Useful set is: $I_1, \dots, I_N, q^2, \zeta \equiv \frac{\rho + \gamma I_1}{1 \gamma^2}$ where I_n are N polynomial rotation invariants of the Hessian matrix ρ_{ij} , $I_1 = \operatorname{Tr} \rho_{ij}$, ..., $I_N = \det[\rho_{ij}]$

(and $I_2 \ldots I_{N-1}$ are built from the minors of orders 2 to N-1)

• Actually, better to use more 'irreducible' combinations J_i, in ND-space

$$J_1 = I_1$$
, $J_{s\geq 2} = I_1^s - \sum_{p=2}^s \frac{(-N)^p C_s^p}{(s-1)C_N^p} I_1^{s-p} I_p$

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Orhogonal polymonial expansion for 2D $\mathscr{P}(\rho, q^2, J_1, J_2)$

Gaussian limit JPDF

$$G_{2D}(\zeta, q^2, J_1, J_2) \,\mathrm{d}\zeta \,\mathrm{d}q^2 \mathrm{d}J_1 \mathrm{d}J_2 = \frac{1}{2\pi} e^{-\frac{1}{2}(\zeta^2 + 2q^2 + J_1^2 + 2J_2)} \mathrm{d}\zeta \,\mathrm{d}q^2 \mathrm{d}J_1 \mathrm{d}J_2$$

serves as the weight for defining the expansion polynomials in

- ζ , $J_1 ([-\infty, \infty]]$, gaussian weight) Hermite
- q^2 , $J_2 ([0, \infty]$, exponential weight) Laguerre.

 $\mathscr{P}_{2\mathrm{D}}(\zeta, q^2, J_1, J_2) = G_{2\mathrm{D}} \times \left[1 + \right]$

$$\sum_{n=3}^{\infty} \sum_{i,j,k,l=0}^{i+2j+k+2l=n} \frac{(-1)^{j+l}}{i!\,j!\,k!\,l!} \left\langle \zeta^{i} q^{2j} J_{1}^{k} J_{2}^{l} \right\rangle_{\rm GC} H_{i}(\zeta) L_{j}(q^{2}) H_{k}(J_{1}) L_{l}(J_{2})$$

This is an expansion to all orders in powers of the field n. Expansion coefficients can be predicted by pertrubation theories

Euler characteristic (genus) as a function of threshold General expression in *ND*

$$\frac{\chi(\nu)}{2} = (-1)^N \int_{\nu}^{\infty} dx \int dq^2 q^{N-1} \delta_D^N(q^2) \int \prod_{s=1}^N dJ_s \mathscr{P}_{ND}(\ldots) I_N$$

can be integrated to give "moment" expansion to all orders

$$\begin{split} \chi(\nu) &= \frac{1}{\sqrt{2\pi}R_*} \exp\left(-\frac{\nu^2}{2}\right) \times \frac{2}{(2\pi)^{N/2}} \left(\frac{\gamma}{\sqrt{N}}\right)^N \left[H_{N-1}(\nu) + \right. \\ &+ \left. \sum_{n=3}^{\infty} \sum_{s=0}^N \gamma^{-s} \sum_{i,j=0}^{i+2j=n-s} \frac{(-N)^{j+s}(N-2)!! L_j^{\left(\frac{N-2}{2}\right)}(0)}{i!(2j+N-2)!!} \left\langle x^i q^{2j} I_s \right\rangle_{GC} H_{i+N-s-1}(\nu) \right] \end{split}$$

(cf first term: Matsubara 1994-2005)

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How non-Gaussianity develops, eq 2D Euler charachteristic

Excitation of Hermite modes of alternating parity



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Summary

"Hermite spectroscopy"



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Redshift space – anisotropic statistics

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Polymonial expansion for \mathcal{P}_{3D} in redshift space

Symmetry:

Rotational around the line of sight (LOS along 3rd coordinate)

Variables:

linear(4) x, x_3 , x_{33} , $J_{1\perp} = x_{11} + x_{22}$ quadratic(3) $q^2 = x_1^2 + x_2^2$, $Q^2 = x_{13}^2 + x_{23}^2$, $J_{2\perp} = (x_{11} - x_{22})^2 + 4x_{12}^2$ cubic(1) $\Upsilon = (x_{13}^2 - x_{23}^2)(x_{11} - x_{22}) + 4x_{12}x_{13}x_{23}$

Gaussian limit JPDF

$$G(x,q_{\perp}^2,x_3\zeta,J_{2\perp},\xi,Q^2,\Upsilon) = \frac{1}{4\pi^3\sqrt{Q^4J_{2\perp}-\Upsilon^2}}e^{-\frac{1}{2}x^2-q_{\perp}^2-\frac{1}{2}x_3^2-\frac{1}{2}\zeta^2-J_{2\perp}-\frac{1}{2}\xi^2-Q^2}$$

Uniformly distributed $\Upsilon \in [-Q^2 \sqrt{J_{2\perp}}, +Q^2 \sqrt{J_{2\perp}}]$ can be integrated over for Minkowski functionals, extrema ...

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Euler characteristic in redshift space

$$\chi_{2+1}(\nu) = \frac{e^{-\nu^2/2}}{8\pi^2} \left[\frac{\sigma_{1\parallel} \sigma_{1\perp}^2}{\sigma^3} H_2(\nu) + \sum_{n=3}^{\infty} \chi_{2+1}^{(n)} \right]$$

with non-Gaussian corrections $\chi^{(n)}_{2+1}$, given, to all orders, by

$$\begin{split} \chi_{2+1}^{(n)}(\nu) &= \frac{\sigma_{2\perp}^2 \sigma_{2\parallel}}{\sigma_{1\perp}^2 \sigma_{1\parallel}} \bigg[\sum_{\sigma_n} \frac{(-1)^{j+m}}{2^m i! \, j! \, m!} H_{i+2}(\nu) \gamma_{\parallel} \gamma_{\perp}^2 \left\langle x^i \, q_{\perp}^{2j} \, x_3^{2m} \right\rangle_{\rm GC} \\ &- \sum_{\sigma_{n-1}} \frac{(-1)^{j+m}}{2^m i! \, j! \, m!} H_{i+1}(\nu) \left\langle x^i \, q_{\perp}^{2j} \, x_3^{2m} \left(\gamma_{\perp}^2 \, x_{33} + 2\gamma_{\perp} \gamma_{\parallel} J_{1\perp} \right) \right\rangle_{\rm GC} \\ &+ \sum_{\sigma_{n-2}} \frac{(-1)^{j+m}}{2^m i! \, j! \, m!} H_i(\nu) \left\langle x^i \, q_{\perp}^{2j} \, x_3^{2m} \left(2\gamma_{\perp} (J_{1\perp} \, x_{33} - \gamma_2 \, Q^2) + \gamma_{\parallel} (J_{1\perp}^2 - J_{2\perp}) \right) \right\rangle_{\rm GC} \\ &- \sum_{\sigma_{n-3}} \frac{(-1)^{j+m}}{2^m i! \, j! \, m!} H_{i-1}(\nu) \left\langle x^i \, q_{\perp}^{2j} \, x_3^{2m} \left(x_{33} (J_{1\perp}^2 - J_{2\perp}) - 2\gamma_2 \left(Q^2 \, J_{1\perp} - \Upsilon \right) \right) \right\rangle_{\rm GC} \bigg] \end{split}$$

Euler characteristics in redshift space



$$\chi_{3D}^{0+1} = \frac{\sigma_{1\parallel}\sigma_{1\perp}^2}{\sigma^3} \frac{e^{-\nu^2/2}}{8\pi^2} \left[H_2(\nu) + \frac{1}{3!} H_5(\nu) \langle x^3 \rangle + H_3(\nu) \left(\langle xq_{\perp}^2 \rangle + \frac{\langle xx_3^2 \rangle}{2} \right) - \frac{H_1(\nu)}{\gamma_{\perp}} \left(\langle J_{1\perp}q_{\perp}^2 \rangle + \langle J_{1\perp}x_3^2 \rangle \right) \right] + \frac{\sigma_{1\parallel}\sigma_{1\perp}^2}{\sigma_{\perp}^2} \left[H_2(\nu) + \frac{1}{3!} H_2(\nu) \langle x^3 \rangle + H_3(\nu) \left(\langle xq_{\perp}^2 \rangle + \frac{\langle xx_3^2 \rangle}{2} \right) - \frac{H_1(\nu)}{\gamma_{\perp}} \left(\langle J_{1\perp}q_{\perp}^2 \rangle + \langle J_{1\perp}x_3^2 \rangle \right) \right] \right]$$

Important anisotropy measure $\beta_{\sigma} = 1 - \frac{\sigma_{1\perp}^2}{2\sigma_{1\parallel}^2} \approx \frac{4}{5}f/b$

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Slicing through redshift space

$$\begin{split} \chi_{2\mathrm{D}}(\nu,\theta) &= \frac{e^{-\nu^2/2}}{(2\pi)^{3/2}} \frac{\sigma_{1\perp}^2}{2\sigma^2} \sqrt{\frac{1-\beta_\sigma \sin^2\theta}{1-\beta_\sigma}} \times \left[H_1(\nu) + \frac{1}{3!}H_4(\nu) \langle x^3 \rangle \right. \\ &+ H_2(\nu) \left(\langle xq_{\perp}^2 \rangle + \frac{1}{2} \frac{\cos^2\theta}{1-\beta_\sigma \sin^2\theta} (\langle xx_3^2 \rangle - \langle xq_{\perp}^2 \rangle) \right) \\ &- \frac{1}{\gamma_{\perp}} \left(\langle q_{\perp}^2 J_{1\perp} \rangle + \frac{1}{2} \frac{\cos^2\theta}{1-\beta_\sigma \sin^2\theta} (\langle x_3^2 J_{1\perp} \rangle - 2 \langle q_{\perp}^2 J_{1\perp} \rangle) \right) + \mathcal{O}(\sigma^2) \right] \end{split}$$

$$\begin{aligned} \mathcal{N}_{2}(\nu,\theta) &= \frac{\sigma_{1\perp}}{2\sqrt{2}\sigma} e^{-\nu^{2}/2} \left(1 + \beta_{\sigma} \frac{\cos^{2}\theta}{4}\right) \times \left[1 + \frac{1}{3!} H_{3}(\nu) \langle x^{3} \rangle \\ &+ \frac{1}{2} H_{1}(\nu) \left(\langle xq_{\perp}^{2} \rangle + \frac{1}{2} \cos^{2}\theta \left(1 + \beta_{\sigma} \frac{3 + 5\sin^{2}\theta}{8}\right) (\langle xx_{3}^{2} \rangle - \langle xq_{\perp}^{2} \rangle) \right) + \mathcal{O}(\sigma^{2}) \right] \end{aligned}$$

$$\mathcal{N}_{1}(\nu,\theta) = \frac{\sigma_{1\perp}}{\sqrt{3}\pi\sigma} e^{-\nu^{2}/2} \sqrt{\frac{1-\beta_{\sigma}\sin^{2}\theta}{1-\beta_{\sigma}}} \times \left[1+\frac{1}{3!}H_{3}(\nu)\langle x^{3}\rangle + \frac{1}{2}H_{1}(\nu)\left(\langle xq_{\perp}^{2}\rangle + \frac{\cos^{2}\theta}{1-\beta_{\sigma}\sin^{2}\theta}\left(\langle xx_{3}^{2}\rangle - \langle xq_{\perp}^{2}\rangle\right)\right) + \mathcal{O}(\sigma^{2})\right]$$

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Invariance of Minkowski functionals

- Minkowski functionals (but also extrema counts) are invariant under any local monotonic transformations x = f(y), if one adjust threshold accordingly, $v_x = f(v_y)$. Which is achived by choosing threshold that gives the same filling factor of the volume above.
- This means that the coefficients in front of every mode in Hermite expansion are invariant separately. This can be shown perturbatively (?).
- Corollary I: combinations of cumulants in every Hermite mode are invariant wrt to any local monothonic bias. (Is it order by order in σ expansion ?)
- Corollary II: If the starting field y is Gaussian, these coefficients are zero (or sum to zero ?)
- Corollary IIa: In anisotropic case, angle dependent section of each coefficient should then vanish separately. In particular

$$\langle xq_{\perp}^2 \rangle = \langle xx_3^2 \rangle \rightarrow \frac{\langle x\nabla_{\perp}x \cdot \nabla_{\perp}x \rangle}{\langle x\nabla_{\parallel} \cdot \nabla_{\parallel} \rangle} = \frac{\sigma_{\perp}^2}{\sigma_{\parallel}^2}$$

- This holds for *f_{NL}* models in leading non Gaussian model explicitly (despite even transformation being local but non-monotonic in this case)
- But here we are talking about local transformation in redshift space.

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Fiducial experiment

- Cut through redshisift space at different angles. For instance, $\theta = 0$ and $\theta = \pi/2$ and measure 2D geometrical statistics as the function of filling factor
- Do Hermite decomposition, determine *H_n* coefficients
- Use the ratio of main Gaussian lines to determine $\beta_{\sigma} = 1 rac{\sigma_{1\perp}^2}{2\sigma_{1\parallel}}$

$$\frac{H_1\big[(\chi_{2D}(\nu_f, \pi/2)\big]}{H_1\big[\chi_{2D}(\nu_f, 0)\big]} = \sqrt{1 - \beta_\sigma} + \mathcal{O}(\sigma^2) \qquad \frac{H_0\big[\mathcal{N}_2(\nu_f, \pi/2)\big]}{H_0\big[\mathcal{N}_2(\nu_f, 0)\big]} = \sqrt{1 - \beta_\sigma} + \mathcal{O}(\sigma^2)$$

Different statistics have different corrections $\mathscr{O}(\sigma^2),$ which can be leveraged for control

 Use angle variable ratio of a secondary to Gaussian line and β_σ to determine normalized cumulant combinations, and their anisotropy

$$\frac{H_2\big[\mathscr{N}_2(\nu_f,\theta)\big]}{H_0\big[\mathscr{N}_2(\nu_f,\theta)\big]} \to \langle xq_{\perp}^2 \rangle, \langle xq_{\perp}^2 \rangle - \langle xx_3^2 \rangle$$

• Using theoretical insight (PT) relate β_{σ} and higher order redshift cumulants to real space $\sigma, f, b \dots$. To evaluate σ 3D statistics have most power.

Using non-Gaussianity and redshift distortions of geometrical measures of the Cosmic Web to recover





Reconstructed β from angular dependence of Minkowski functionals in 2D slices

Reconstructed σ from non-Gaussian corrections of Minkowski functionals

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Conclusions

- Redshift space distortions are not just nuisance, but a source of additional information about cosmology of the Universe.
- Redshift space analysis is in principle capable of mining more information than real space analysis. For that one needs to leverage the anisotropic properties of redshift space.
- Generalizing previous work, we have developed complete theory of Minkowski functionals in the bulk and on slices of anisotropic and mildly non-Gaussian fields.
- At mildly non-linear scales of cosmological structure formation, non-Gaussian and redshift effects can be, to some extend, disentangled.