# Bayesian reconstruction of the cosmic DM flow

Florian Führer IAP, Paris

#### with Guilhem Lavaux









The Aquila consortium

## **Cosmological large-scale structure**

- Small initial conditions, approximate scale free and Gaussian
- Today still linear on large scales, non-linear on small scales
- Use to test of Dark Energy, Dark Matter, Gravity, Neutrino mass ...
- DM not directly observable: use galaxies as tracers, lensing, ...



#### Large-scale structure reconstruction

- Typically one tries to measure summary statistics like density Power Spectrum
- A complete treatment requires a reconstruction of the density/velocity field itself
- A fully Bayesian treatment requires obtaining the high dimensional posterior
- A lot of progress on density reconstruction

Jasche, Kitaura 2010, Jasche, Wandelt 2013, Modi et. al. 2018 Jasche, Lavaux 2018 and many more

• In this talk: Direct reconstruction of the DM velocity field See also Lavaux 2013

## The DM velocity field

The DM velocity can be inferred from density Leclergc et. al. 2017

Direct reconstruction, allows to test Euler equation

$$\partial_t \delta + \nabla \cdot \left( (1+\delta) \mathbf{v} \right) = 0$$
$$\partial_t \mathbf{v} + H \mathbf{v} + \mathbf{v} \cdot \nabla \mathbf{v} = -\nabla \Phi$$

Measured galaxy redshifts combine  $\frac{(1+z)}{(1+\bar{z})} = (1 + \mathbf{v} \cdot \mathbf{n})$ 

Need to incorporate distance measurements  $\bar{z} = d_L(\bar{z})$ 

**Velocity** bias

Galaxies are no perfect tracers of the DM density

$$\delta(z, \mathbf{x}) = b \, n_{gal}(z, \mathbf{x})$$

Galaxies do not perfectly trace the DM velocity neither

Error-model:

Assume  $\epsilon_{\rm NL}$  to be gaussian  $\sigma_{\rm NL}^2$  can be different for different galaxies  $\rightarrow$  classify galaxies into types with different  $\rightarrow$  determine self-consistently

$$\mathbf{v} \cdot \mathbf{n} = \frac{z - z}{1 + \bar{z}} + \epsilon_{\rm NL}$$
Stochastic contribution

#### Statistical model and likelihood





### Statistical model and likelihood



### Hamilton Monte Carlo

Reformulate as Hamiltonian particle system, with auxiliary variables (momentum) P Neal 2011

$$H = \frac{1}{2}\mathbf{p} \cdot \mathbf{M} \cdot \mathbf{p} - \log\left(P\left(\theta|\mathcal{D}\right)\right)$$

Samples obtained by solving Hamiltonian E.o.M.

Jasche, Kitaura 2010

$$\dot{\theta} = \mathbf{M}^{-1} \cdot \mathbf{p} \qquad \dot{\mathbf{p}} = -\partial_{\theta} \log \left( P\left(\theta | \mathcal{D} \right) \right)$$

Hamiltonian Monte Carlo is well suited to sample from high dimensional distributions

- Travel large distances in parameter space
- High acceptance rate
- Use gradient information

#### 6dF distance data



- About 9000 galaxies
- Only the southern sky

### **Redshift selection**

Redshift likelihood gaussian, but typically surveys are cut a at maximal redshift

$$P(z|z_o) \propto \mathcal{N}(z_o|z)\theta(z_o - z_{cut})\frac{1}{Z(z, z_{cut})}$$
Posterior not gaussian,  
due normalization
$$Z(z, z_{cut}) \propto \operatorname{erfc}\left(\frac{z_{cut} - z}{\sigma_z}\right)$$

Not taking into account would lead  $\sum_{z_o} \frac{|z_{cut}|}{|z_o|}$  to in falling of galaxies on large scales

#### **Convergence** analysis

#### Preliminary



### **Error-model** distribution

### Preliminary



### Radial velocity field

#### Preliminary



### Density fluctuations in 6dF

#### Radial density profile of density fluctuations

From Lavaux & Jasche 2018



### **Conclusion and Outlook**

- Direct reconstruction of 3D velocity field
  - 9000 galaxies from 6df in southern hemisphere
  - In future: Add 2000 Spitzer galaxies to analysis
- Implement in BORG framework Jasche, Wandelt 2013
  - Joined reconstruction with density
  - Reconstruction of the non-linear field
- How can we use the velocity field
  - Test gravity?
  - Implications for Hubble constant measurements?