5 December, TeVPA 2019

Probing the history of dark sector interactions

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Dark sector interactions

- Experimental searches for particle DM
- No experimental search for DE
 - DM candidates motivated by theory (WIMPs, axions, sterile neutrinos, etc)
 - DM-DE interactions?
- Probing dark sector interaction in the lab a non-starter
- However...

... the question of DM-DE interaction can (only?) be investigated from cosmological observations

Interacting DE models

- DM and DE generically coupled
- Dynamical DE model + coupling have been studied for over two decades
 - elaborate models; weak observational constraints
 - alleviates 'coincidence problem'
- Recent work focussed on interaction itself
 - minimal extension to ΛCDM
 - fluid description of DE

Interaction in fluid approximation

• Conservation of total energy-momentum given by

$$\nabla_{\mu}(T_{b}^{\mu\nu} + T_{c}^{\mu\nu} + T_{de}^{\mu\nu} + T_{\gamma}^{\mu\nu} + T_{\nu}^{\mu\nu} + \cdots) = 0$$

• Relax the Λ CDM assumption $\nabla_{\mu}T_{c}^{\mu\nu} = 0$, $\nabla_{\mu}T_{de}^{\mu\nu} = 0$:

$$\nabla_{\mu} T_{c}^{\mu\nu} = -Q^{\nu}$$
$$\nabla_{\mu} T_{de}^{\mu\nu} = +Q^{\nu}$$

but still have total conservation of EM

- Decompose wrt CDM 4-velocity $Q^{\mu}(t, \mathbf{x}) = Q(t)u_c^{\mu} + f^{\mu}$
- Restrict to energy exchange only ($f^{\mu} = 0$):

$$Q^{\mu}(t,\mathbf{x}) = \mathbf{Q}(t)u_{c}^{\mu}$$

• Q typically chosen for simplicity but no theoretical basis

This work: Probe DM-DE interaction using a datadriven, model-independent approach

Reconstruction method

- Take agnostic approach: let data decide best form of ${\cal Q}$
- Define dimensionless interaction strength

$$q(a) := \frac{Q(a)}{(H\rho_{de})}$$

• q(a) has infinite DoF. We reconstruct a discretized representation

$$q(a) = \sum_{i=1}^{n_{\text{bins}}} q_i T_i(a) \implies \text{define a vector } \mathbf{q} = (q_1, \dots, q_{n_{\text{bins}}})^T$$
$$T_i(a) = \begin{cases} 1, & a_{i-1} \le a \le a_i \\ 0, & \text{otherwise} \end{cases}$$

Smoothing prior

Bayes' theorem: $p(\mathbf{q} | \text{data}) \propto p(\text{data} | \mathbf{q}) p(\mathbf{q})$

- Too much freedom, degeneracies abound
- Restrict to subspace of smooth reconstructions

Assign informative prior to ${\boldsymbol{q}}$

$$p(\mathbf{q}) = \frac{1}{\sqrt{\det 2\pi C}} \exp\left(-\frac{1}{2}(\mathbf{q}^{fid} - \mathbf{q})^T C_{prior}^{-1}(\mathbf{q}^{fid} - \mathbf{q})\right)$$
$$\mathbf{q}^{fid} = (q_1^{fid}, \dots, q_{p_{rior}}^{fid})^T$$

- reconstruction regresses to \mathbf{q}^{fid} in the absence of data
- key to smoothness lies in C_{prior}

Specifying correlations

- Bin-bin correlations $C_{ij} = \langle (q_i \bar{q}_i)(q_j \bar{q}_j) \rangle$
- Discretize:

$$q(a) \to q_i = \int_0^1 da \, W_i(a) \, q(a)$$
$$C_{ij} = \int_0^1 da \, W_i(a) \int_0^1 da' \, W_j(a') \, \xi(|a-a'|)$$

where

$$\xi(\left|a-a'\right|) := \left\langle (q(a)-q^{fid}(a))(q(a')-q^{fid}(a')) \right\rangle$$

 a_2

 a_{i}

 a_3

 a_i

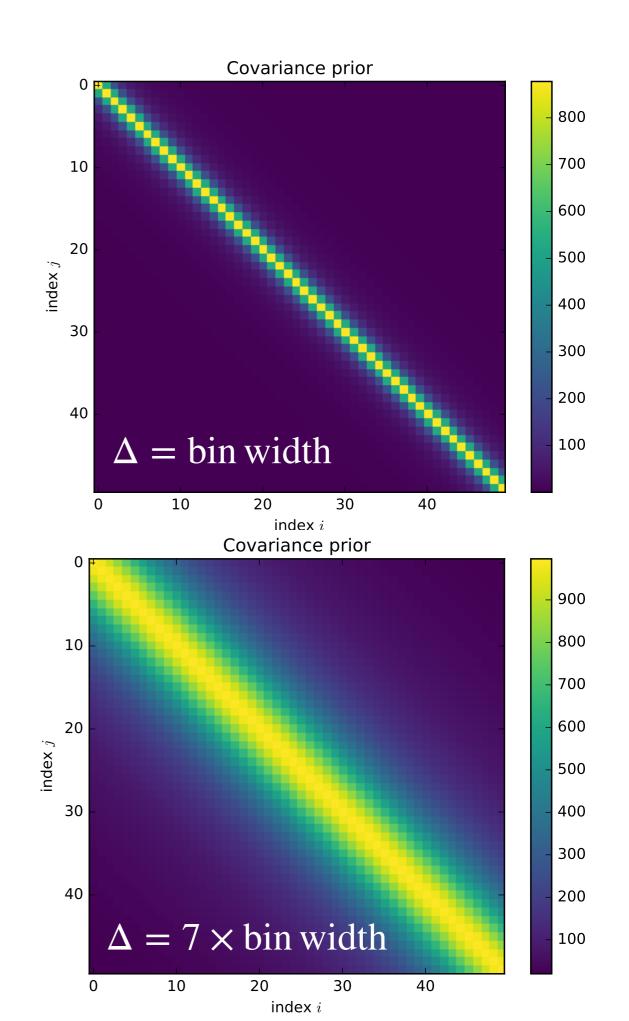
 $a'_{2} a'_{3}$

• To compute C_{ij} sum all pairwise correlations between bin i and bin j

 For smoothness we want bins to be positively correlated. We adopt the correlation function (astro-ph/0510293)

$$\xi(|a - a'|) = \frac{\sigma^2}{1 + (|a - a'|/\Delta)^2}$$

- σ controls strength of correlations
- Δ controls characteristic scale of correlations



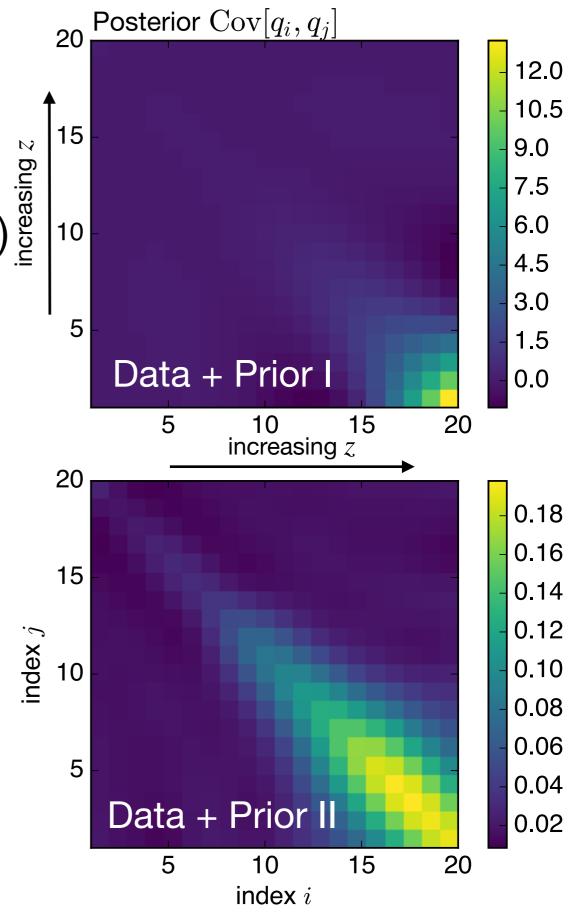
Specifying fiducial model

Consider two fiducial models:

Prior I: $\mathbf{q}^{fid} = A\mathbf{q}$ (running average)

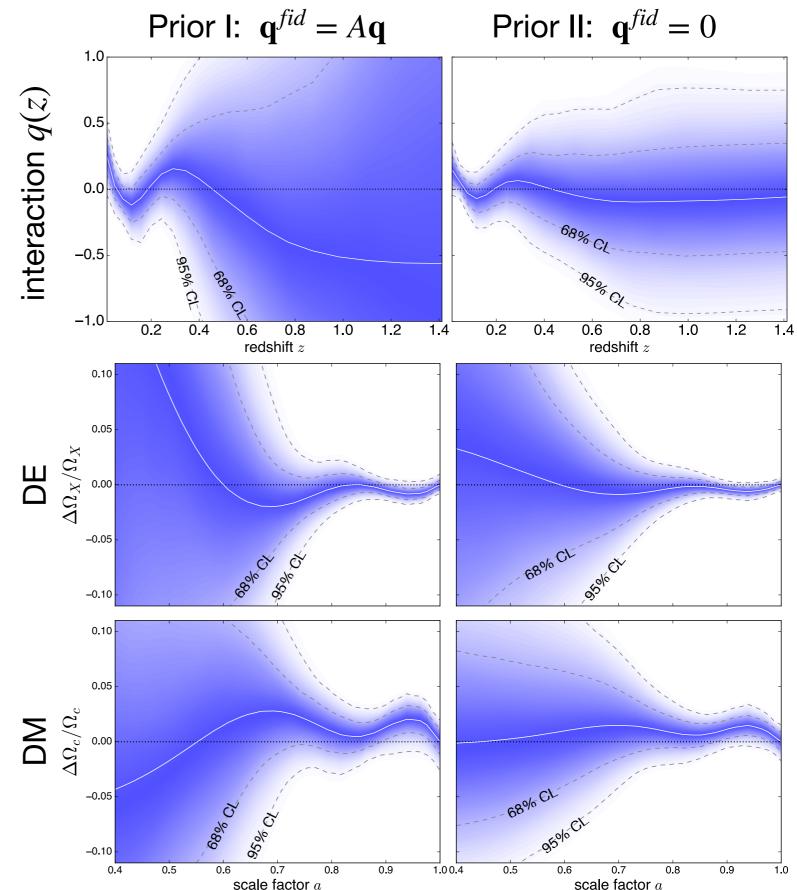
Prior II: $\mathbf{q}^{fid} = 0$ (conservative)

• Data more informative at low-z, uncertainties in q_i increase with i



Full reconstruction

- Data = BAO+CMB+RSD+SN+ H(z)
- Interaction consistent with zero at z > 0.2
- Slight feature at $z \leq 0.2$ robust to two different priors
- Consistent with zero but lots of redundancy. Want optimal description of q(z)



Karhunen-Loève analysis

$$q(a) = \sum_{i} q_{i} T_{i}(a) = \sum_{i} \alpha_{i} e_{i}(a)$$
KL modes

- Generalizes aspects of PCA
- KL identifies features that are well-determined. Noise dominated features can be filtered out
- Principal component analysis (PCA) not valid since we have non-trivial prior information

- does not distinguish between prior and data

KL signal-to-noise modes

$$q(a) = \sum_{i} q_i T_i(a) = \sum_{i} \alpha_i e_i(a)$$

 What's important is the information gained from likelihood (data)

 $Cov[\alpha_m, \alpha_n]$ is diagonal, $\{\alpha_i\}$ are uncorrelated variables

$$C_{prior} \mathbf{e}_{a} = \lambda_{a} C_{post} \mathbf{e}_{a}$$

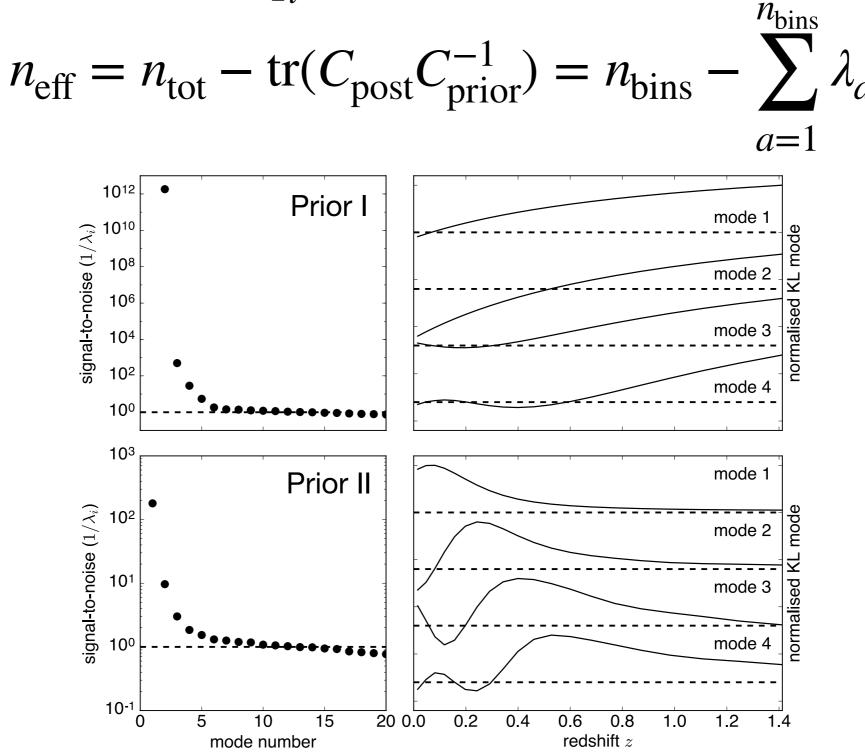
$$(C_{post} \mathbf{e}_{a} = \lambda_{a} \mathbf{e}_{a} \text{ for PCA})$$

• Unlike PCA $\{\mathbf{e}_1, \mathbf{e}_2, ..., \mathbf{e}_{n_{\mathrm{bins}}}\}$ are not orthogonal set. Instead

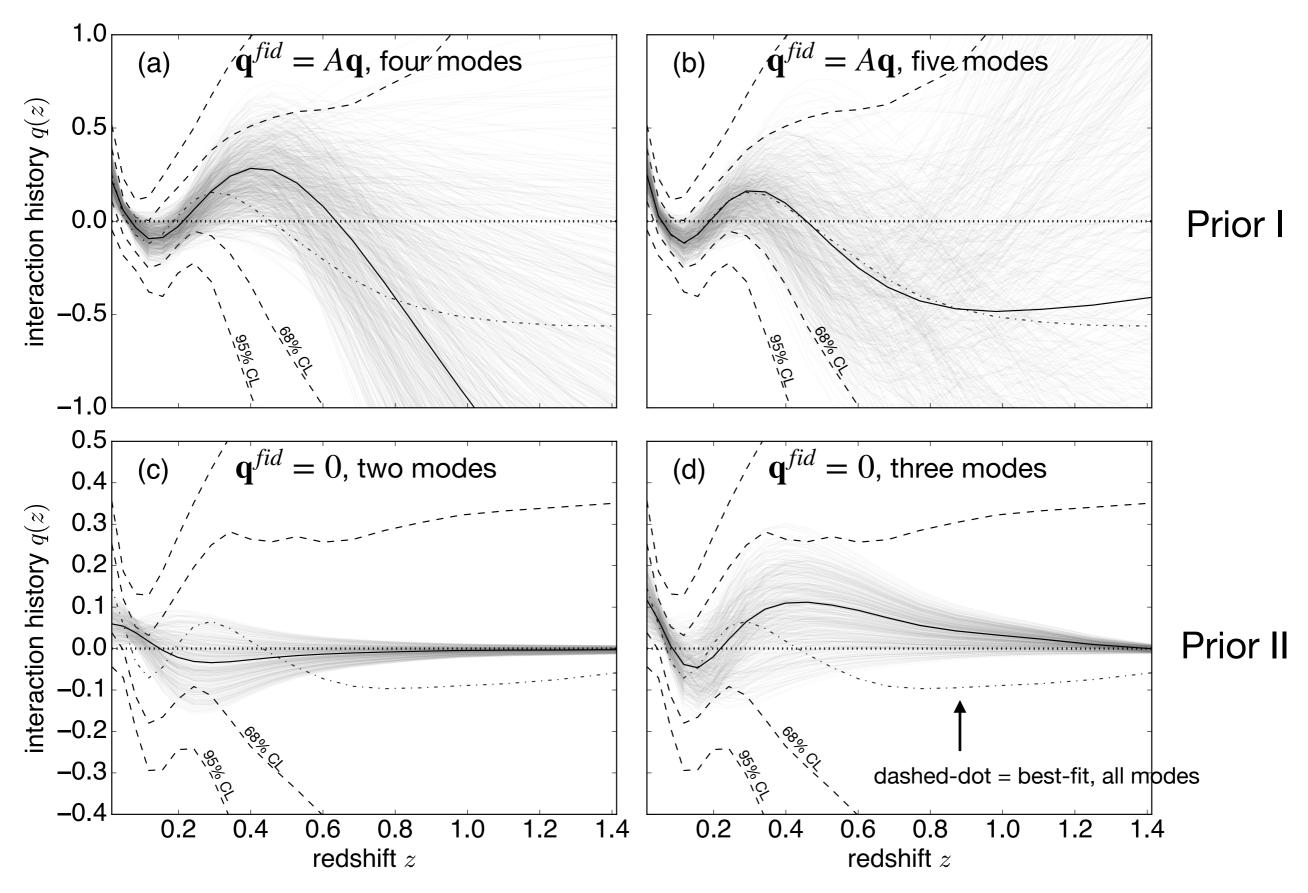
$$\mathbf{e}_{a}^{T} C_{post} \mathbf{e}_{b} = \delta_{ab}$$
$$\mathbf{e}_{a}^{T} C_{prior} \mathbf{e}_{b} = \lambda_{a} \delta_{ab} \qquad 1/\lambda_{a} \text{ is S/N of } a^{th} \text{ mode}$$

How many KL modes to retain?

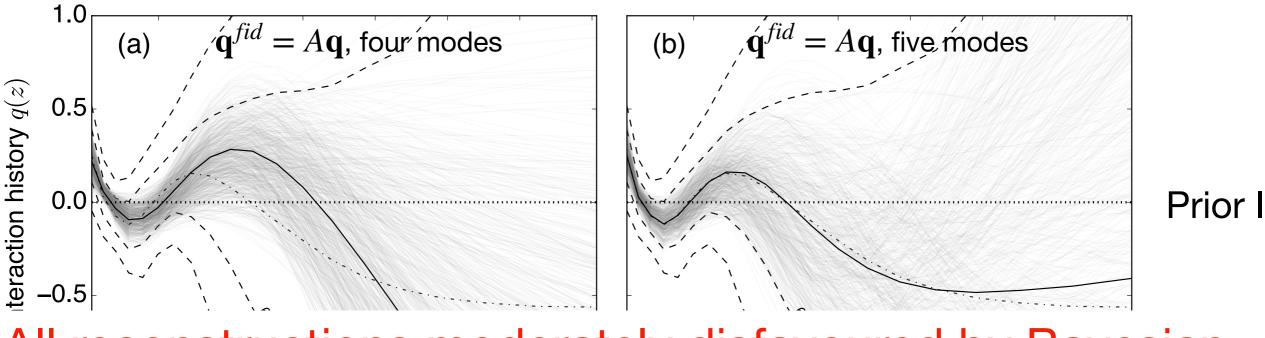
• Effective number of q_i parameters



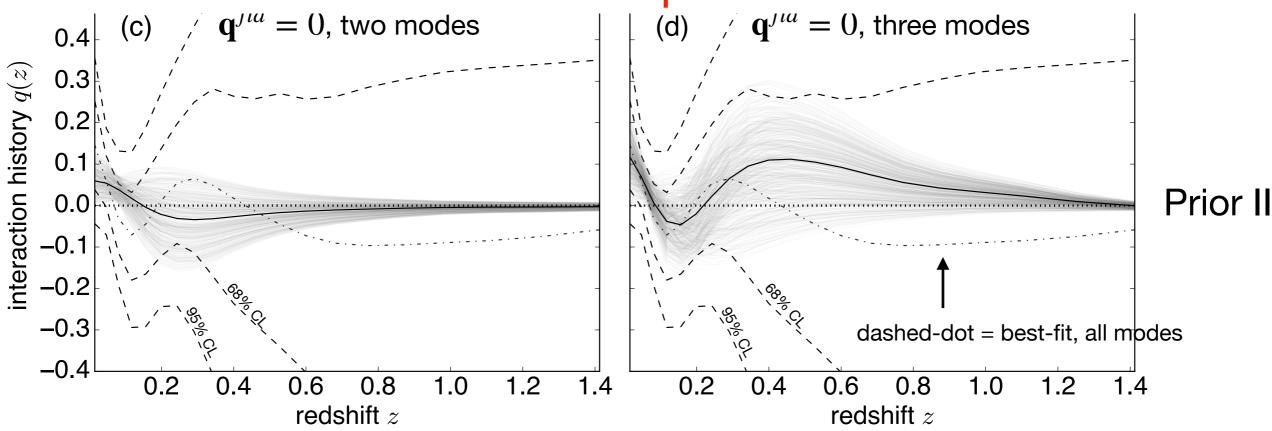
KL reconstruction



KL reconstruction



All reconstructions moderately disfavoured by Bayesian model comparison

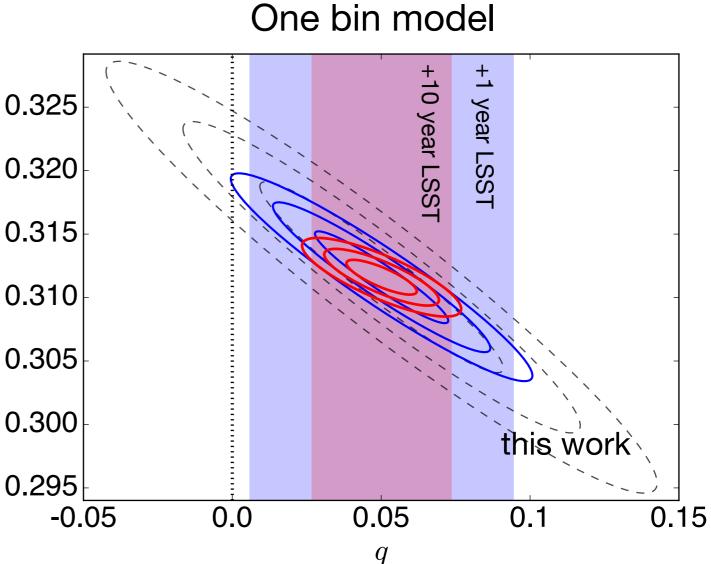


Future prospects

• Fisher matrix $F_{ij} = -\left. \left\langle \frac{\partial^2 \ln \mathscr{L}}{\partial \theta^i \partial \theta^j} \right\rangle \right|_{\substack{\theta_i = \theta_i^{ML} \\ \text{expect from DESI (RSD+BAO) } 0.305 \\ \text{and LSST SNIa?} \\ 0.300}$

- > 3σ constraints on interaction possible with 1 year LSST sample
 - improvements largely come from RSD

$$F_{ij} = F_{ij}^{\text{LSST-SN}} + F_{ij}^{\text{DESI-RSD}} + F_{ij}^{\text{DESI-BAO}} + F_{ij}^{\text{data}}$$



Summary

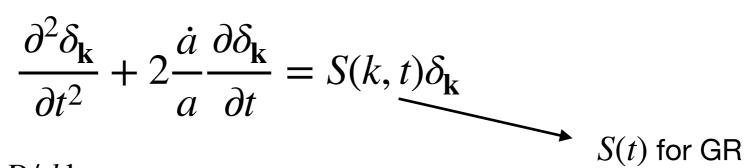
- 1. No evidence for late-time interaction in the dark sector
 - Targeted search for DM-DE interaction shows slight $\lesssim 2\sigma$ deviations from $\Lambda {\rm CDM}$ at z < 0.2
- 2. KL analysis can be used to isolate data-sensitive modes
 - needed for penalised reconstruction
 - One KL mode well-constrained with current data
- 3. Λ CDM remains best model in all model comparison tests when range of data considered

Thank you

Extra slides

Interaction or MG?

- In linear perturbation theory $\delta(t, \mathbf{x}) = D(t) \delta_i(\mathbf{x})$ i.e. grows uniformly with growth factor D(t)
 - relative height of peaks and troughs of density field does not change with time (selfsimilar)
 - consequence of inverse-square law. More generally



• Growth rate $f = d \ln D/d \ln a$

 $f(a) \approx \Omega_m(a)^{\gamma}, \quad \gamma \approx 0.55 \text{ for } \Lambda \text{CDM}$

- background expansion history fully determines f(a) (non-local)
- Measurements that differ from $\gamma \approx 0.55$ taken as sign of gravity theory *not* given by GR
 - provides a framework for investigating MG
- Interaction complicates this picture
 - can still have GR and $\gamma \neq 0.55$ with interaction