## Sparsity Weak Lensing Mass Reconstruction

## ABSTRACT

## 1. INTRODUCTION

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#### 2. THE PROBLEM

## 2.1. Density Contrast to Shear

The lensing distortion  $\kappa$  at a position  $\vec{x} = (\vec{\theta}, \chi_s)$  caused by the foreground inhomogeneous density distribution is

$$\kappa(\vec{\theta}, \chi_s) = \frac{3H_0^2 \Omega_M}{2c^2} \int_0^{\chi_s} d\chi_l \frac{\chi_l \chi_{sl}}{\chi_s} \frac{\delta(\chi_l \vec{\theta}, \chi_l)}{a(\chi_l)}$$
(1)

(Leonard et al. 2014). Substitute angular distance D with redshift z and we have

$$\kappa(\vec{\theta}, z_s) = \int_0^{z_s} dz_l \delta_c^{-1}(z_l, z_s) \delta(\vec{\theta}, z_l). \tag{2}$$

we term  $\delta_c(z_l, z_s)$  as lensing critical density contrast and define it as

$$\delta_c^{-1}(z_l, z_s) = \begin{cases} \frac{3H_0 \Omega_M}{2c} \frac{\chi_l \chi_{sl} (1 + z_l)}{\chi_s E(z_l)} & (z_s > z_l), \\ 0 & (z_s \le z_l). \end{cases}$$
(3)

As reported in Kaiser & Squires (1993), the shear distortion is

$$\gamma(\vec{\theta}, z_s) = \int D(\vec{\theta} - \vec{\theta'}) \kappa(\vec{\theta'}, z_s) d^2\theta, \tag{4}$$

where

$$D(\vec{\theta}) = -\frac{1}{\pi} (\theta_1 - i\theta_2)^{-2},$$
  

$$\gamma(\vec{\theta}) = \gamma_1(\vec{\theta}) + i\gamma_2(\vec{\theta}).$$
(5)

Combine equation (2) with equation (4) and we have

$$\gamma(\vec{\theta}, z_s) = \int_0^{z_s} dz_l \frac{\delta_c(z_l, +\infty)}{\delta_c(z_l, z_s)} \int d^2\theta' D(\vec{\theta} - \vec{\theta'}) \frac{\delta(\vec{\theta'}, z_l)}{\delta_c(z_l, +\infty)}.$$
(6)

#### 2.2. Photo-z Uncertainty

Since the redshifts of source galaxies are estimated with a limited number of photometric bands, the estimated redshift of a galaxy suffer from large uncertainty. The probability distribution function (PDF) of a galaxies at position  $(\vec{\theta}, z_s)$  is denoted as  $\mathcal{P}(z|z_s)$ . Note that, in

order to simplify the future calculation, we assume the variation of the PDF across the transverse plane is small and neglect the dependency on the transverse position.

Taking the uncertainty of redshift into account, equation (6) changes to

$$\gamma(\vec{\theta}, z_s) = \int_0^{z_s} dz_l P(z_l, z_s) \gamma(\vec{\theta}, z_l, +\infty),$$

$$\gamma(\vec{\theta}, z_l, +\infty) = \int d^2 \theta' Q(\vec{\theta} - \vec{\theta'}, z_l) \delta(\vec{\theta'}, z_l),$$
(7)

where  $\gamma(\vec{\theta}, z_l, +\infty)$  represents shear signal at infinite redshift  $(z_s = +\infty)$  caused by density contrast at  $z_l$ , and  $P(z_l, z_s)$  is the convolution kernel along line-of-sight and  $Q(\vec{\theta}, \vec{\theta}', z_l)$  is the convolution kernel along on the transverse plane. These kernels are defined as

$$P(z_{l}, z_{s}) = \int \frac{\delta_{c}(z_{l}, +\infty)}{\delta_{c}(z_{l}, z)} \mathcal{P}(z|z_{s}) dz,$$

$$Q(\vec{\theta}, \vec{\theta'}, z_{l}) = \frac{D(\vec{\theta} - \vec{\theta'})}{\delta_{c}(z_{l}, +\infty)}.$$
(8)

## 2.3. Smoothing

Since the observed galaxies have random irregular (unequally-spaced) distribution, it is necessary to smooth the shear signal in the observation. The smoothing is expressed as follows

$$\gamma_S(\vec{\theta}, z) = \frac{\sum_i W(\vec{\theta} - \vec{\theta_i}, z - z_i) e_i}{\sum_i W(\vec{\theta} - \vec{\theta_i}, z - z_i) R_i},\tag{9}$$

where  $W(\vec{\theta}, z)$  is a 3-D smoothing kernel.  $e_i$ ,  $R_i$ ,  $z_i$  and  $\theta_i$  are the ellipticity, response, reshift, transverse position of the 'i-th' galaxy.  $W(\vec{\theta}, z)$  can be decomposed into a transverse component  $W_T(\vec{\theta})$  and a line-of-sight component  $W_{\times}(z)$ 

$$W(\vec{\theta}, z) = W_T(\vec{\theta}) W_{\times}(z). \tag{10}$$

In this paper, we set

$$W_T(\vec{\theta}) = \frac{1}{2\pi\beta^2} \exp(-\frac{|\vec{\theta}|}{2\beta^2}),$$

$$W_{\times}(z) = \begin{cases} 1/\Delta z & (|z| < \Delta z/2), \\ 0 & else. \end{cases}$$
(11)

With the assumption that the density of response R and the density of galaxy number vary slowly on the smoothing scale, the smoothed shear is

$$\gamma_S(\vec{\theta}_j, z_j) = \int d^2\theta_s dz_s W(\vec{\theta}_j - \vec{\theta}_s, z_j - z_s) \gamma(\vec{\theta}_s, z_s) \tag{12}$$

Substitute equation (7) into equation (12)

$$\gamma_S(\vec{\theta}_j, z_j) = \int_0^{z_j} dz_l P_S(z_l, z_j) \gamma_S(\vec{\theta}_j, z_l, +\infty),$$

$$\gamma_S(\vec{\theta}_j, z_l, +\infty) = \int d^2 \theta' Q_S(\vec{\theta}_j, \vec{\theta}', z_l) \delta(\vec{\theta}', z_l),$$
(13)

and

$$P_S(z_l, z_j) = \int dz_s W_{\times}(z_j - z_s) \int \frac{\delta_c(z_l, +\infty)}{\delta_c(z_l, z)} \mathcal{P}(z|z_s) dz,$$

$$Q_S(\vec{\theta}, \vec{\theta'}, z_l) = \int d^2 \theta'' W_T(\vec{\theta} - \vec{\theta''}) \frac{D(\vec{\theta''} - \vec{\theta'})}{\delta_c(z_l, +\infty)}.$$
(14)

## 2.4. Mask and Noise

In real observations, the influence of mask and noise should also be taken into account. The final observed shear is

$$\gamma_o(\vec{\theta}, z) = M(\vec{\theta}, z)\gamma(\vec{\theta}, z) + \epsilon(\vec{\theta}, z), \tag{15}$$

where  $\epsilon$  represents noise typically caused by random shape of intrinsic galaxies and such noise is modeled as white Gaussian noise.  $M(\vec{\theta}, z_s)$  is the mask function.

We can write equation (15) into

$$\gamma_o = \mathbf{MPQ}\delta + \epsilon, \tag{16}$$

where M, P and Q represent the functional operators M, P and Q, respectively.

## 3. THE METHOD

## 3.1. Dictionary

Since many N-body simulations have shown the dark matter to be largely distributed in halos connected by filaments, we assume that the density contrast field can be decomposed into multi-scaled NFW halo (Navarro et al. 1997) and point mass at different positions

$$\delta(\vec{x}) = \sum_{s=0}^{N_s} \int d^3 x' \phi_s(\vec{x} - \vec{x}') \alpha_s(\vec{x}')$$
 (17)

where  $\phi_0$  is 3-D Dirac delta function representing point mass

$$\phi_0(\vec{x}) = \delta_D(\theta_1)\delta_D(\theta_2)\delta_D(z). \tag{18}$$

Since the scale of halo is much less than the reachable redshift resolution, we neglect the depth of halo on the line-of-sight direction as suggested by (Leonard et al. 2014). On the transverse plane, the NFW atom is modeled with the surface density profile of NFW halo with scale  $\theta_s$  and truncation radius  $c\theta_s$  (Takada & Jain 2003), where c is generally known as concentration of NFW halo. Whereas, the NFW atom is modeled with Dirac delta function on the line-of-sight direction

$$\phi_s(\vec{x}) = \frac{f}{2\pi\theta_s^2} F(|\vec{\theta}|/\theta_s) \delta_D(z),$$

$$(s = 1..N_s)$$
(19)

where

$$F(x) = \begin{cases} -\frac{\sqrt{c^2 - x^2}}{(1 - x^2)(1 + c)} + \frac{1}{(1 - x^2)^{3/2}} \operatorname{arccosh}(\frac{x^2 + c}{x(1 + c)}) & (x < 1), \\ \frac{\sqrt{c^2 - 1}}{3(1 + c)}(1 + \frac{1}{c + 1}) & (x = 1), \\ -\frac{\sqrt{c^2 - x^2}}{(1 - x^2)(1 + c)} + \frac{1}{(x^2 - 1)^{3/2}} \operatorname{arccos}(\frac{x^2 + c}{x(1 + c)}) & (1 < x \le c), \\ 0 & (x > c). \end{cases}$$

$$(20)$$

and 
$$f = 1/[\ln(1+c) - c/(1+c)].$$

 $\alpha_s$  is the corresponding decomposition coefficient of the density contrast field onto our basis atom. The total

coefficients set is denoted as 
$$\alpha=\begin{pmatrix}\alpha_0\\\alpha_1\\\dots\\\alpha_{N_s}\end{pmatrix}$$
 , and the total

dictionary is 
$$\Phi = \left(\phi_0, \phi_1, ..., \phi_{N_s}\right)$$

We pixelize the parameter space into a  $N_{\theta} \times N_{\theta} \times N_{l}$  grid.  $N_{\theta}$  is the number of pixels on the two dimensions of the transverse plane and  $N_{l}$  is the number of pixels on the line-of-sight direction for the lenses. Similarly,  $\gamma_{o}$  is pixelized onto a  $N_{\theta} \times N_{\theta} \times N_{s}$  grid, where  $N_{s}$  is the number of pixel on the line-of-sight direction for the sources. The pixel size for the two dimensions on the transverse plane is denoted as  $\Delta\theta$  and the pixel size for the line-of-sight direction is denoted as  $\Delta z$ . Equation (16) changes to

$$\gamma = \mathbf{MPQ}\Phi\alpha + \epsilon. \tag{21}$$

### 3.2. Loss Function with Constrains

The loss function is defined as

$$L = \|\gamma - \mathbf{MPQ}\Phi\alpha\|_{2}^{2} + \lambda \|\alpha\|_{1}^{1} + \tau TSV(\alpha_{0}), \quad (22)$$

where  $\|\gamma - \mathbf{MPQ\Phi}\alpha\|_2^2$  is the normal chi-square,  $\|\alpha^{N,P}\|_1^1$  is the sparsity constrain, and  $\mathrm{TSV}(\alpha^P)$  is the Total Squared Variation (TSV) constrain.  $\|\bullet\|_1$  and  $\|\bullet\|_2$  represent  $l_1$  norm and  $l_2$  norm, respectively. The

TSV operator is defined as

$$TSV(\alpha_0) = \sum_{ijk} \{ (\alpha_0[i, j, k] - \alpha_0[i+1, j, k])^2 + (\alpha_0[i, j, k] - \alpha_0[i, j+1, k])^2 + (\alpha_0[i, j, k] - \alpha_0[i, j, k+1])^2 \},$$
(23)

where  $i = 1..N_{\theta}$ ,  $j = 1..N_{\theta}$ , are the pixel indexes for the two dimensions on the transverse plane,  $k = 1..N_l$  is the pixel index on the line-of-sight direction.

We define

$$f(\alpha) = \|\gamma - \mathbf{MPQ}\Phi\alpha\|_{2}^{2} + \tau TSV(\alpha^{P})$$
$$= \sum_{lm} A_{lm}\alpha_{l}\alpha_{m} + \sum_{l} B_{l}\alpha_{l} + C,$$
(24)

where l, m go over the indexes of (i, j, k, s).

3.3. *FISTA* 

The basic FISTA algorithms is

$$X_m^{(n+1)} = \operatorname{ST}_{\lambda}(\alpha_m - \frac{\mu}{2A_{mm}} \frac{\partial f}{\partial \alpha_m})|_{\alpha_m^{(n)}}, \qquad (25)$$

where

$$ST_{\lambda}(x) = sign(x) \max(|x| - \lambda, 0)$$
 (26)

is the soft thresholding function.

$$t^{(n+1)} = \frac{1 + \sqrt{1 + 4(t^{(n)})^2}}{2},$$

$$\alpha_m^{(n+1)} = X_m^{(n+1)} + \left[\frac{t^{(n)} - 1}{t^{(n+1)}} (X_m^{(n+1)} - X_m^{(n)})\right],$$
(27)

with  $t^{(0)} = 1$  and  $X^{(0)} = \alpha^{(0)}$ .

4. SIMULATION

4.1. Postage Stamp

4.1.1. One Halo

4.1.2. Multiple Halo

4.2. Mock catalog

5. RESULTS

6. SUMMARY

#### REFERENCES

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# APPENDIX

## A. GENERAL LINEAR LASSO

Note that the loss function defined in equation (22) into

$$L(\alpha) = f(\alpha) + \lambda \|\alpha\|_{1}^{1}, \tag{A1}$$

where

$$f(\vec{\alpha}) = \vec{\alpha}^T \mathbf{A} \vec{\alpha} + \vec{B} \cdot \vec{\alpha} + C. \tag{A2}$$

Using Einstein notation, we have

$$L(\alpha_i) = A_{ij}\alpha_i\alpha_j + B_i\alpha_i + C + \lambda_i|\alpha_i|. \tag{A3}$$

Set the initial parameter vector is  $\vec{\alpha}^{(0)}$ . We focus on one specific parameter with index  $\mu$  and fix other parameters with indexes  $i \neq \mu$ 

$$L(\alpha_{\mu}|\alpha_{i\neq\mu}^{(0)}) = A_{\mu\mu}\alpha_{\mu}^{2} + (A_{i\mu}\alpha_{i}^{(0)} + A_{\mu i}\alpha_{i}^{(0)} + B_{\mu})\alpha_{\mu} + \lambda_{\mu}|\alpha_{\mu}| + B_{i}\alpha_{i}^{(0)} + \lambda_{i}|\alpha_{i}^{(0)}| + C$$
(A4)

reach its minimum at  $\mathrm{ST}_{\lambda}(\alpha_{\mu} - \frac{\partial_{\mu} f(\alpha_{\mu} | \alpha_{i \neq \mu}^{(0)})}{2A_{\mu\mu}})|_{\alpha_{\mu}^{(0)}}$ , where  $\mathrm{ST}_{\lambda}(x) = \mathrm{sign}(x) \max(|x| - \lambda, 0)$  is the soft thresholding function. Then the parameter can be updated

$$\alpha_{\mu}^{(1)} \leftarrow \mathrm{ST}_{\lambda}(\alpha_{\mu} - \frac{\partial_{\mu} f(\alpha_{\mu} | \alpha_{i \neq \mu}^{(0)})}{2A_{\mu\mu}})|_{\alpha_{\mu}^{(0)}}.\tag{A5}$$

Other parameters are subsequently updated in the same way. The minimum of the loss function can be reached by iterating the parameter update.