1 ADMM details

We first re-parameterize $\phi_j = y - \theta_j$ so the problem is

minimize
$$\rho_{\tau}(\phi) + \lambda ||D^{(k)}(y - \phi)||_1$$
 (1)

We further divide ϕ order to solve smaller problems: Defining

$$\phi_1 = (\phi_{11}, \phi_{12}) \tag{2}$$

$$\phi_2 = (\phi_{21}, \phi_{22}, \phi_{23}) \tag{3}$$

$$\phi_3 = (\phi_{31}, \phi_{32}) \tag{4}$$

$$\phi = (\phi_{11}, \phi_{12} = \phi_{21}, \phi_{22}, \phi_{23} = \phi_{31}, \phi_{32}) \tag{5}$$

(6)

Dividing y similarly, the problem then becomes

minimize
$$\sum_{i=1}^{3} \rho_{\tau}(\phi_{i}) + \lambda ||D^{(k)}(y_{i} - \phi_{i})||_{1}$$
 (7)

subject to:
$$\phi_{12} = \phi_{21}$$
, $\phi_{23} = \phi_{31}$ (8)

(9)

We can further simplify by defining

$$\overline{\phi} = (\phi_{11}, \frac{\phi_{12} + \phi_{21}}{2}, \phi_{22}, \frac{\phi_{23} + \phi_{31}}{2}, \phi_{32})$$
 (10)

$$\overline{\phi_1} = (\phi_{11}, \frac{\phi_{12} + \phi_{21}}{2}) \tag{11}$$

$$\overline{\phi_2} = (\frac{\phi_{12} + \phi_{21}}{2}, \phi_{22}, \frac{\phi_{23} + \phi_{31}}{2}) \tag{12}$$

$$\overline{\phi_3} = (\frac{\phi_{23} + \phi_{31}}{2}, \phi_{32}) \tag{13}$$

so the problem becomes

minimize
$$\sum_{i=1}^{3} \rho_{\tau}(\phi_i) + \lambda ||D^{(k)}(y_i - \phi_i)||_1$$
 (14)

subject to:
$$\phi_i = \overline{\phi_i}$$
 (15)

(16)

The augmented Lagrangian for this problem is

$$L_{\gamma}(\phi_1, \phi_2, \phi_3, \overline{\phi_1}, \overline{\phi_2}, \overline{\phi_3}, \omega) = \tag{17}$$

$$\sum_{i=1}^{3} \rho_{\tau}(\phi_{i}) + \lambda ||D^{(k)}(y_{i} - \phi_{i})||_{1} + \omega_{i}^{T}(\phi_{i} - \overline{\phi_{i}}) + \frac{\gamma}{2} ||\phi_{i} - \overline{\phi_{i}}||_{2}^{2}$$
(18)

The ADMM updates are then given by

$$\phi_i^{k+1} = \arg\min_{\phi_i} \rho_{\tau}(\phi_i) + \lambda ||D^{(k)}(y_i - \phi_i)||_1 + \omega_i^{kT}(\phi_i - \overline{\phi_i}^k) + \frac{\gamma}{2} ||\phi_i - \overline{\phi_i}^k||_2^2$$
 (19)

$$\omega_i^{k+1} = \omega_i^k + \gamma(\phi_i^{k+1} - \overline{\phi_i}^{k+1}) \tag{20}$$

The ϕ_i updates can be obtained using a quadratic program solver such as Gurobi and can be obtained in parallel.

2 Simulation Metrics

Figure 1: F1 score by threshold, data size, and method (1 is best 0 is worst).

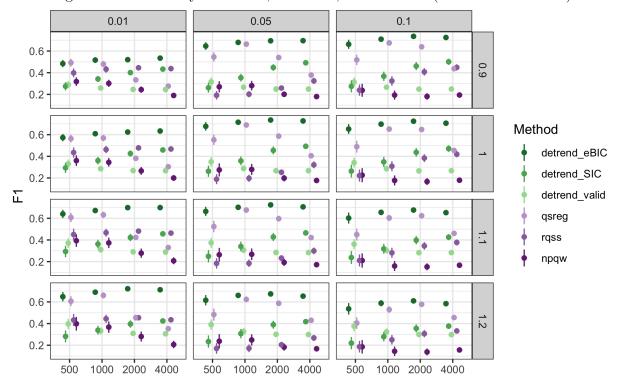


Figure 2: Precision by threshold, data size, and method (true positive over true positives + false positives).

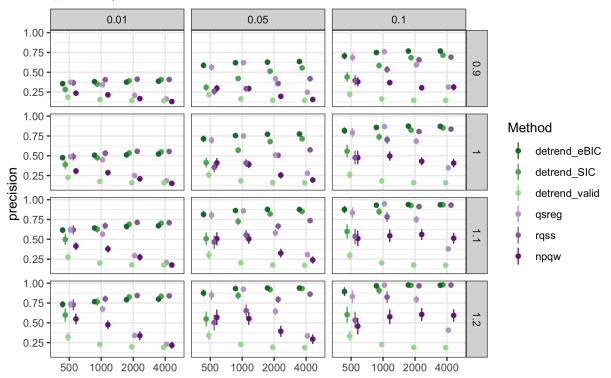


Figure 3: Recall by threshold, data size, and method (true positive over true positives + false negatives).

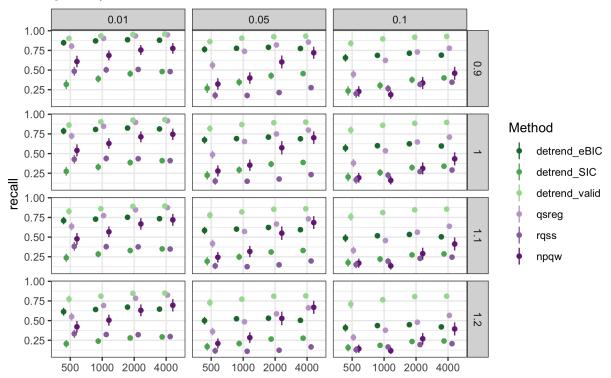


Figure 4: Miss-classification rates by threshold, data size, and method, values above the upper limit (npqw) not shown.

