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# PROPERTIES AND ITERATIVE METHODS FOR THE ELASTIC NET WITH $\ell_p$ -NORM ERRORS

LILING WEI\* AND HONG-KUN XU\*\*

### \*School of Science, Hangzhou Dianzi University, Hangzhou, 310018, China E-mail: wll.1225@foxmail.com

\*\*School of Science, Hangzhou Dianzi University, Hangzhou, 310018, China E-mail: xuhk@hdu.edu.cn (Corresponding author)

Abstract. The *p*-elastic net (p-EN) with  $1 is introduced to recover a sparse signal <math>x \in \mathbb{R}^n$  from m (< n) linear measurements with noise. The *p*-EN, which extends the elastic net of Zou and Hastie [23] and was implicitly suggested by Tropp [16], amounts to minimizing the objective function  $(1/p) ||Ax - b||_p^p + \lambda ||x||_1 + (\mu/2) ||x||_2^2$  over  $x \in \mathbb{R}^n$ , where A is the measurement matrix, b is the observation, and  $\lambda > 0$ ,  $\mu > 0$  are regularization parameters. Some basic geometric properties of the *p*-EN such as how the solution curve of the minimization depends on the parameters  $\lambda$  and  $\mu$  are investigated. Moreover, iterative algorithms such as the proximal-gradient algorithm and the Frank-Wolfe algorithm are studied for solving the *p*-EN.

Key Words and Phrases: Lasso, compressed sensing, elastic net,  $\ell_p$ -norm error, proximal gradient, Frank-Wolfe.

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