# A coordinate gradient descent method for nonsmooth separable minimization

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February 6, 2014

### **Problem**

minimize 
$$\underbrace{f(x)}_{\text{smooth}} + \underbrace{g(x)}_{\text{nonsmooth}}$$

### **Problem**

$$\underset{\mathsf{smooth}}{\text{minimize}} \quad \underbrace{f(x)}_{\mathsf{smooth}} \quad + \quad \lambda ||x||_{1}$$

#### **Problem**

minimize 
$$\underbrace{f(x)}_{\text{smooth}} + \lambda ||x||_1$$

State-of-the-art approach for sparse inverse covariance estimation, QUIC algorithm (Hsieh et al. 2011)

 $minimize f(x) + \overline{\lambda ||x||_1}$ 

minimize 
$$f(x) + \lambda ||x||_1$$

#### Repeat

1. Form the second-order Taylor expansion

$$\hat{f}(x+\Delta) = f(x) + \nabla f(x)^T \Delta + \frac{1}{2} \Delta^T \nabla^2 f(x) \Delta$$

minimize 
$$f(x) + \lambda ||x||_1$$

#### Repeat

1. Form the second-order Taylor expansion

$$\hat{f}(x + \Delta) = f(x) + \nabla f(x)^T \Delta + \frac{1}{2} \Delta^T \nabla^2 f(x) \Delta$$

2. Solve for the regularized Newton step

$$d = \arg\min_{\Delta} \hat{f}(x + \Delta) + \lambda ||x + \Delta||_1$$

minimize 
$$f(x) + \lambda ||x||_1$$

#### Repeat

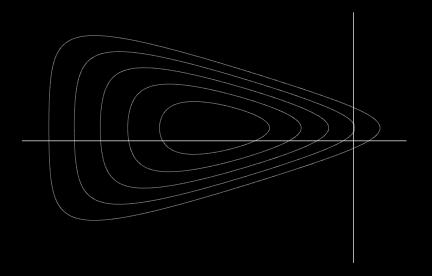
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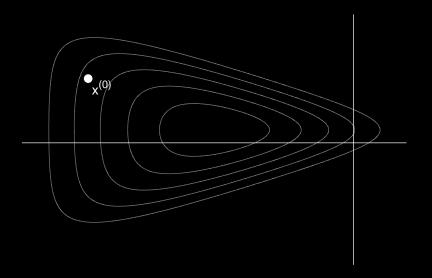
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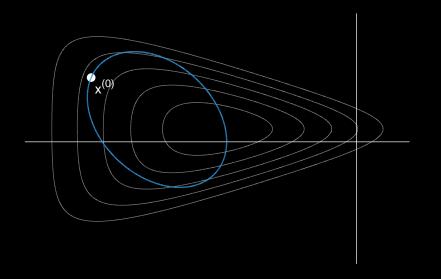
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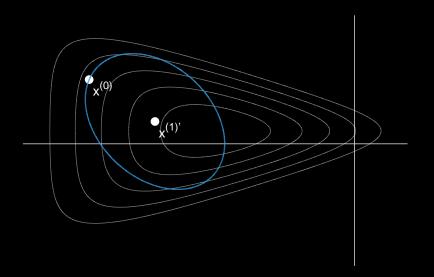
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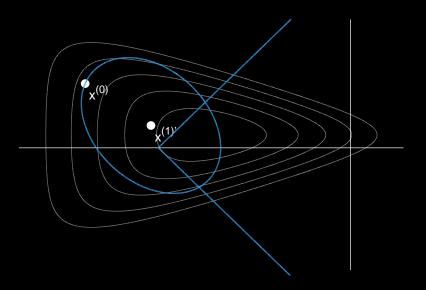
3. Update x using backtracking line search

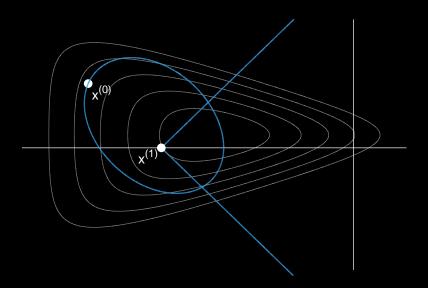


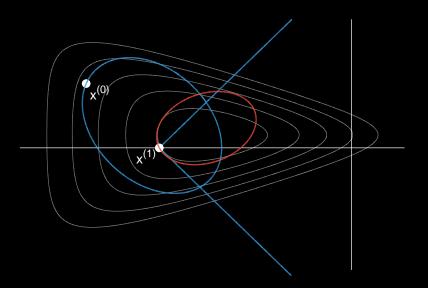


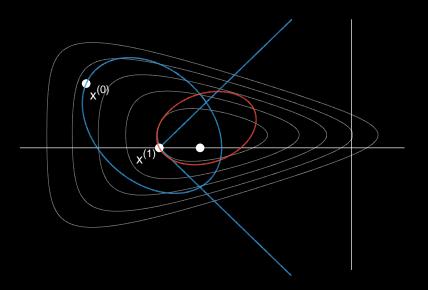


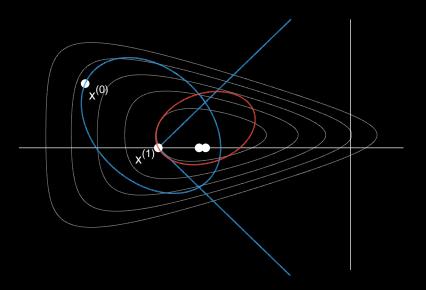












### Computational complexity of inner loop

$$d = \arg\min_{\Delta} \hat{f}(x + \Delta) + \lambda ||x + \Delta||_1$$

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Exploit sparsity using a coordinate descent active set method

1. Form active set by including coordinate i if

$$|\nabla f(x)_i| > \lambda \text{ or } x_i \neq 0$$

2. Iteratively minimize, done in closed form

# Theoretical analysis

### Theoretical analysis

It works.

#### Theoretical analysis (for real)

1. The objective F=f+g is nonincreasing,  $F>-\infty$ 

2. 
$$F(x^{k+1}) - F(x^k) \le \sigma \alpha^k \Delta^k \le 0$$

3.  $\inf_k \alpha^k > 0$ 

4. 
$$\Delta^k \to 0, d^k \to 0$$
 as  $k \to \infty$ 

5.  $||x^k - x^*|| \to 0$