

11.1 INTERIOR-POINT METHODS FOR CONVEX QUADRATIC PROGRAM

- The form of the problem solved:

$$\begin{aligned} \min_x \quad & q(x) = \frac{1}{2}x^T Gx + x^T c \\ \text{subject to} \quad & Ax \geq b, \end{aligned}$$

$$A = [a_i]_{i \in \mathcal{I}}, \quad b = [b_i]_{i \in \mathcal{I}}, \quad \mathcal{I} = \{1, 2, \dots, m\}.$$

Optimality Conditions

- In original form:

$$Gx - A^T \lambda + c = 0,$$

$$Ax - b \geq 0,$$

$$(Ax - b)_i \lambda_i = 0, \quad i = 1, 2, \dots, m,$$

$$\lambda \geq 0.$$

- With slacks:

$$Gx - A^T \lambda + c = 0,$$

$$Ax - y - b = 0,$$

$$y_i \lambda_i = 0, \quad i = 1, 2, \dots, m,$$

$$(y, \lambda) \geq 0.$$

Idea: define an “interior” path to the solution

- Define the perturbed KKT conditions as a nonlinear system:

- Solve su $F(x, y, \lambda; \sigma \mu) = \begin{bmatrix} Gx - A^T \lambda + c \\ Ax - y - b \\ \mathcal{Y} \Lambda e - \sigma \mu e \end{bmatrix} = 0,$

$$F(x(\mu), y(\mu), \lambda(\mu); \mu, \sigma) = 0; y_i(\mu) \lambda_i(\mu) = \mu \sigma > 0$$

$$\mu \rightarrow 0 \Rightarrow (x(\mu), y(\mu), \lambda(\mu)) \rightarrow (x^*, y^*, \lambda^*) \text{ satisfies KKT}$$

How to solve it ?

- Solution: apply Newton's method for fixed μ :

$$\begin{bmatrix} G & 0 & -A^T \\ A & -I & 0 \\ 0 & \Lambda & \mathcal{Y} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \\ \Delta \lambda \end{bmatrix} = \begin{bmatrix} -r_d \\ -r_p \\ -\Lambda \mathcal{Y} e + \sigma \mu e \end{bmatrix},$$

- New iter

$$r_d = Gx - A^T \lambda + c, \quad r_p = Ax - y - b.$$

- Enforce: $(x^+, y^+, \lambda^+) = (x, y, \lambda) + \alpha(\Delta x, \Delta y, \Delta \lambda),$

$$(y^+, \lambda^+) > 0;$$

How to PRACTICALLY solve it?

- Eliminate the slacks:

$$\begin{bmatrix} G & -A^T \\ A & \Lambda^{-1}\mathcal{Y} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta \lambda \end{bmatrix} = \begin{bmatrix} -r_d \\ -r_p + (-y + \sigma \mu \Lambda^{-1}e) \end{bmatrix}.$$

- Eliminate the multipliers (note, the 22 block is diagonal and invertible)

$$(G + A^T \mathcal{Y}^{-1} \Lambda A) \Delta x = -r_d + A^T \mathcal{Y}^{-1} \Lambda [-r_p - y + \sigma \mu \Lambda^{-1}e],$$

- Solve (e.g by Cholesky if G is PD and modified Cholesky if not).

- The QP has the SAME structure as an EQP and can use projected CG:

$$\begin{bmatrix} G & 0 & -A^T \\ 0 & \mathcal{Y}^{-1}\Lambda & I \\ A & -I & 0 \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \\ \Delta \lambda \end{bmatrix} = \begin{bmatrix} -r_d \\ -\Lambda e + \sigma \mu \mathcal{Y}^{-1} e \\ -r_p \end{bmatrix},$$

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How to choose the step?

- We need to enforce positivity.
- A typical approach:

$$\alpha_{\tau}^{\text{pri}} = \max\{\alpha \in (0, 1] : y + \alpha \Delta y \geq (1 - \tau)y\}, \quad \alpha = \min(\alpha_{\tau}^{\text{pri}}, \alpha_{\tau}^{\text{dual}}),$$
$$\alpha_{\tau}^{\text{dual}} = \max\{\alpha \in (0, 1] : \lambda + \alpha \Delta \lambda \geq (1 - \tau)\lambda\};$$

- Here tau is a user defined parameter (say 0.1 – 0.01).

A practical primal-dual method

- First, compute an affine scaling step (that is, drive to solution and not to center) $\sigma = 0$. This allows us to move faster. Denote it by

$$(\Delta x^{\text{aff}}, \Delta y^{\text{aff}}, \Delta \lambda^{\text{aff}})$$

- Then, move towards the center to make sure that, taking a Newton from this point to the center.

$$\begin{bmatrix} G & 0 & -A^T \\ A & -I & 0 \\ 0 & \Lambda & \mathcal{Y} \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \\ \Delta \lambda \end{bmatrix} = \begin{bmatrix} -r_d \\ -r_p \\ -\Lambda \mathcal{Y} e - \Delta \Lambda^{\text{aff}} \Delta \mathcal{Y}^{\text{aff}} e + \sigma \mu e \end{bmatrix}$$

A practical primal-dual algorithm

Algorithm 16.4 (Predictor-Corrector Algorithm for QP).

Compute (x_0, y_0, λ_0) with $(y_0, \lambda_0) > 0$;

for $k = 0, 1, 2, \dots$

Set $(x, y, \lambda) = (x_k, y_k, \lambda_k)$ and solve (16.58) with $\sigma = 0$ for
 $(\Delta x^{\text{aff}}, \Delta y^{\text{aff}}, \Delta \lambda^{\text{aff}})$;

Calculate $\mu = y^T \lambda / m$;

Calculate $\hat{\alpha}_{\text{aff}} = \max\{\alpha \in (0, 1] \mid (y, \lambda) + \alpha(\Delta y^{\text{aff}}, \Delta \lambda^{\text{aff}}) \geq 0\}$;

Calculate $\mu_{\text{aff}} = (y + \hat{\alpha}_{\text{aff}} \Delta y^{\text{aff}})^T (\lambda + \hat{\alpha}_{\text{aff}} \Delta \lambda^{\text{aff}}) / m$;

Set centering parameter to $\sigma = (\mu_{\text{aff}} / \mu)^3$;

Solve (16.67) for $(\Delta x, \Delta y, \Delta \lambda)$;

Choose $\tau_k \in (0, 1)$ and set $\hat{\alpha} = \min(\alpha_{\tau_k}^{\text{pri}}, \alpha_{\tau_k}^{\text{dual}})$ (see (16.66));

Set $(x_{k+1}, y_{k+1}, \lambda_{k+1}) = (x_k, y_k, \lambda_k) + \hat{\alpha}(\Delta x, \Delta y, \Delta \lambda)$;

end (for)

11.2 INTERIOR-POINT METHODS

- Same idea as in the case of the interior-point method for QP.
- Create a path that is interior with respect to the Lagrange multipliers and the slacks that depends on a smoothing parameter μ .
- Drive μ to 0.

Interior –point, “smoothing” path

- Formulation (with slacks) :

$$\begin{aligned}
 & \min_{x,s} f(x) \\
 & \text{subject to} \quad c_E(x) = 0, \\
 & \quad \quad \quad c_I(x) - s = 0, \\
 & \quad \quad \quad s \geq 0.
 \end{aligned}$$

- Interior-point (smoothing path; $\mu=0$: KKT)

$$\begin{aligned}
 \nabla f(x) - A_E^T(x)y - A_I^T(x)z &= 0, & c_E(x) &= 0, \\
 Sz - \mu e &= 0, & c_I(x) - |s| &= 0,
 \end{aligned}$$

Barrier interpretation

- The nonlinear equation is the same as the KKT point of the barrier function:

$$\begin{aligned} \min_{x,s} \quad & f(x) - \mu \sum_{i=1}^m \log s_i \\ \text{subject to} \quad & c_E(x) = 0, \\ & c_I(x) - s = 0, \end{aligned}$$

- Linearization for fixed μ :

$$\begin{bmatrix} \nabla_{xx}^2 \mathcal{L} & 0 & -A_E^T(x) & -A_I^T(x) \\ 0 & Z & 0 & S \\ A_E(x) & 0 & 0 & 0 \\ A_I(x) & -I & 0 & 0 \end{bmatrix} \begin{bmatrix} p_x \\ p_s \\ p_y \\ p_z \end{bmatrix} = - \begin{bmatrix} \nabla f(x) - A_E^T(x)y - A_I^T(x)z \\ Sz - \mu e \\ c_E(x) \\ c_I(x) - s \end{bmatrix},$$

$$\mathcal{L}(x, s, y, z) = f(x) - y^T c_E(x) - z^T (c_I(x) - s).$$

- The new iteration:

$$x^+ = x + \alpha_s^{\max} p_x, \quad s^+ = s + \alpha_s^{\max} p_s,$$

- Where: $y^+ = y + \alpha_z^{\max} p_y, \quad z^+ = z + \alpha_z^{\max} p_z,$

- And, $\alpha_s^{\max} = \max\{\alpha \in (0, 1] : s + \alpha p_s \geq (1 - \tau)s\},$
 $\alpha_z^{\max} = \max\{\alpha \in (0, 1] : z + \alpha p_z \geq (1 - \tau)z\},$

$$\tau = 0.99 - 0.995$$

How do I measure progress?

- Merit function:

$$E(x, s, y, z; \mu) = \max \left\{ \|\nabla f(x) - A_E(x)^T y - A_I(x)^T z\|, \|S z - \mu e\|, \right. \\ \left. \|c_E(x)\|, \|c_I(x) - s\| \right\},$$

- I try to decrease it as much as I can.

Basic Interior-Point Algorithm

Algorithm 19.1 (Basic Interior-Point Algorithm).

Choose x_0 and $s_0 > 0$, and compute initial values for the multipliers y_0 and $z_0 > 0$.
Select an initial barrier parameter $\mu_0 > 0$ and parameters $\sigma, \tau \in (0, 1)$. Set $k \leftarrow 0$.

repeat until a stopping test for the nonlinear program (19.1) is satisfied

 repeat until $E(x_k, s_k, y_k, z_k; \mu_k) \leq \mu_k$

 Solve (19.6) to obtain the search direction $p = (p_x, p_s, p_y, p_z)$;

 Compute $\alpha_s^{\max}, \alpha_z^{\max}$ using (19.9);

 Compute $(x_{k+1}, s_{k+1}, y_{k+1}, z_{k+1})$ using (19.8);

 Set $\mu_{k+1} \leftarrow \mu_k$ and $k \leftarrow k + 1$;

 end

 Choose $\mu_k \in (0, \sigma \mu_k)$;

end

How to solve the linear system

- Rewriting the Newton Direction:

$$\begin{bmatrix} \nabla_{xx}^2 \mathcal{L} & 0 & A_E^T(x) & A_I^T(x) \\ 0 & \Sigma & 0 & -I \\ A_E(x) & 0 & 0 & 0 \\ A_I(x) & -I & 0 & 0 \end{bmatrix} \begin{bmatrix} p_x \\ p_s \\ -p_y \\ -p_z \end{bmatrix} = - \begin{bmatrix} \nabla f(x) - A_E^T(x)y - A_I^T(x)z \\ z - \mu S^{-1}e \\ c_E(x) \\ c_I(x) - s \end{bmatrix}$$

- Can use indefinite factorization LDLT.
- Or, projected CG (since it is in saddle-point form)

$$\Sigma = S^{-1}Z.$$

- Or, we can eliminate p_s and use LDLT

$$\begin{bmatrix} \nabla_{xx}^2 \mathcal{L} & A_E^T(x) & A_I^T(x) \\ A_E(x) & 0 & 0 \\ A_I(x) & 0 & -\Sigma^{-1} \end{bmatrix} \begin{bmatrix} p_x \\ -p_y \\ -p_z \end{bmatrix} = - \begin{bmatrix} \nabla f(x) - A_E^T(x)y - A_I^T(x)z \\ c_E(x) \\ c_I(x) - \mu Z^{-1}e \end{bmatrix}$$

- And even p_z :

$$\begin{bmatrix} \nabla_{xx}^2 \mathcal{L} + A_I^T \Sigma A_I & A_E^T(x) \\ A_E(x) & 0 \end{bmatrix}$$

How do we deal with nonconvexity and non-LICQ?

- Regularization

$$\begin{bmatrix} \nabla_{xx}^2 \mathcal{L} + \delta I & 0 & A_E(x)^T & A_I(x)^T \\ 0 & \Sigma & 0 & -I \\ A_E(x) & 0 & -\gamma I & 0 \\ A_I(x) & -I & 0 & 0 \end{bmatrix}.$$

- Choose delta so that signature of the matrix corresponds to positive definiteness of reduced matrix:
- For signature, can use LDLT

$$(n + m, l + m, 0)$$

But, how do I know how far to go in a direction?

- Backtracking search for merit function (based on barrier interpretation) :

$$\phi_v(x, s) = f(x) - \mu \sum_{i=1}^m \log s_i + v \|c_E(x)\| + v \|c_I(x) - s\|,$$

- Direction $\alpha_s \in (0, \alpha_s^{\max}]$, $\alpha_z \in (0, \alpha_z^{\max}]$,

$$\frac{\partial}{\partial p} \|c(x)\| = \frac{\partial}{\partial p} \sqrt{c(x)^T c(x)} = \begin{cases} \frac{c(x)}{\|c(x)\|} \nabla c(x) p & c(x) \neq 0 \\ \frac{\nabla c(x) p}{\|\nabla c(x) p\|} \nabla c(x) p & c(x) = 0, \nabla c(x) p \neq 0 \\ 0 & \text{otherwise} \end{cases}$$

How do we update barrier parameter?

- Decrease of barrier (example):

$$\mu_{k+1} = \sigma_k \mu_k, \quad \text{with } \sigma_k \in (0, 1).$$

$$\sigma_k = 0.1 \min \left(0.05 \frac{1 - \xi_k}{\xi_k}, 2 \right)^3, \quad \text{where } \xi_k = \frac{\min_i [s_k]_i [z_k]_i}{(s^k)^T z^k / m}.$$

- Step update:

$$\begin{aligned} x^+ &= x + \alpha_s p_x, & s^+ &= s + \alpha_s p_s, \\ y^+ &= y + \alpha_z p_y, & z^+ &= z + \alpha_z p_z. \end{aligned}$$

A practical interior-point algorithm

Algorithm 19.2 (Line Search Interior-Point Algorithm).

Choose x_0 and $s_0 > 0$, and compute initial values for the multipliers y_0 and $z_0 > 0$. If a quasi-Newton approach is used, choose an $n \times n$ symmetric and positive definite initial matrix B_0 . Select an initial barrier parameter $\mu > 0$, parameters $\eta, \sigma \in (0, 1)$, and tolerances ϵ_μ and ϵ_{TOL} . Set $k \leftarrow 0$.

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repeat until  $E(x_k, s_k, y_k, z_k; 0) \leq \epsilon_{\text{TOL}}$ 
  repeat until  $E(x_k, s_k, y_k, z_k; \mu) \leq \epsilon_\mu$ 
    Compute the primal-dual direction  $p = (p_x, p_s, p_y, p_z)$  from
      (19.12), where the coefficient matrix is modified as in
      (19.25), if necessary;
    Compute  $\alpha_s^{\max}, \alpha_z^{\max}$  using (19.9); Set  $p_w = (p_x, p_s)$ ;
    Compute step lengths  $\alpha_s, \alpha_z$  satisfying both (19.27) and
       $\phi_v(x_k + \alpha_s p_x, s_k + \alpha_s p_s) \leq \phi_v(x_k, s_k) + \eta \alpha_s D\phi_v(x_k, s_k; p_w)$ ;
    Compute  $(x_{k+1}, s_{k+1}, y_{k+1}, z_{k+1})$  using (19.28);
    if a quasi-Newton approach is used
      update the approximation  $B_k$ ;
    Set  $k \leftarrow k + 1$ ;
  end
  Set  $\mu \leftarrow \sigma \mu$  and update  $\epsilon_\mu$ ;
end

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