

Lecture 16

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### 1 Last Time:

**Theorem 1 (KM Theorem)** Suppose  $N : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is 1-Lipschitz continuous, i.e.  $(\forall x \in \mathbb{R}^n)(\forall y \in \mathbb{R}^n) \|Nx - Ny\| \leq \|x - y\|$ , that  $\text{Fix}(N) \neq \emptyset$ , and  $\lambda \in (0, 1)$ . Then, given any  $z^0 \in \mathbb{R}^n$  the sequence  $\{z^k\}_{k \in \mathbb{N}}$  generated by the KM iteration

$$z^{k+1} = N_\lambda z^k = (1 - \lambda)z^k + \lambda N z^k$$

converges to an element of  $\text{Fix}(N)$ .

### 2 The Method of Alternating Projections (MAP)

Suppose

$$x^* \in \text{argmin} \{c^T x \mid Ax = b, x \geq 0\} \quad \text{and} \quad (y^*, s^*) \in \text{argmax} \{b^T y \mid A^T y + s = c, s \geq 0\}.$$

Using strong duality, these inclusions are equivalent to

$$Ax^* = b; \quad A^T y^* + s = c; \quad c^T x^* - b^T y^* = 0; \quad x^* \geq 0; \quad s^* \geq 0.$$

Define the set  $C_1$  as the solutions to

$$\begin{bmatrix} A & 0 & 0 \\ 0 & A^T & 1 \\ c^T & -b^T & 0 \end{bmatrix} \begin{bmatrix} x^* \\ y^* \\ s^* \end{bmatrix} = \begin{bmatrix} b \\ c \\ 0 \end{bmatrix}$$

and the set  $C_2 = \{(x, y, s) \in \mathbb{R}^{m+2n} \mid x, s \geq 0\}$ . We have just shown that LPs can actually be cast as a **feasibility problem**:

**Theorem 2** The pair  $(x^*, y^*)$  is primal-dual optimal if, and only if, there exists  $s^* \in \mathbb{R}_{\geq 0}$  such that  $(x^*, y^*, s^*) \in C_1 \cap C_2$ .

Now, let's take a step back and consider two closed, convex sets  $C_1, C_2 \subseteq \mathbb{R}^n$ . Let's solve,  $x \in C_1 \cap C_2$  by forming an operator  $N : \mathbb{R}^n \rightarrow \mathbb{R}^n$  with fixed points  $C_1 \cap C_2$ . To apply the KM theorem, the operator  $N$  must be 1-Lipschitz continuous.

**Definition 1** We call a 1-Lipschitz mapping  $N : \mathbb{R}^n \rightarrow \mathbb{R}^n$  **nonexpansive**.

We will often find the following identity useful:

**Lemma 3** For all  $a, b \in \mathbb{R}^n$  and  $\lambda \in \mathbb{R}$ , we have

$$\|(1 - \lambda)a + \lambda b\|^2 = (1 - \lambda)\|a\|^2 + \lambda\|b\|^2 - \lambda(1 - \lambda)\|a - b\|^2.$$

Before we construct the operator  $N$ , we prove a Lemma which shows that projection mappings satisfy a property slightly stronger than nonexpansiveness.

**Lemma 4** Let  $C \subseteq \mathbb{R}^n$  be a closed convex set. Then  $(\forall x \in \mathbb{R}^n)(\forall y \in \mathbb{R}^n)$

$$\|P_C(x) - P_C(y)\|^2 \leq \|x - y\|^2 - \|(x - P_C(x)) - (y - P_C(y))\|, \quad (\text{firm non-expansiveness})$$

In particular,  $P_C$  and  $2P_C - I$  are nonexpansive.

**Proof:** Recall that  $x - P_C(x) \in N_C(P_C(x))$  and  $y - P_C(y) \in N_C(P_C(y))$ , so

$$\langle x - P_C(x), P_C(y) - P_C(x) \rangle \leq 0 \quad \text{and} \quad \langle y - P_C(y), P_C(x) - P_C(y) \rangle \leq 0.$$

Add these inequalities to get

$$\begin{aligned} 0 &\geq \langle (x - P_C(x)) - (y - P_C(y)), P_C(y) - P_C(x) \rangle, \\ &= \frac{1}{2} (-\|x - y\|^2 + \|(x - P_C(x)) - (y - P_C(y))\|^2 + \|P_C(x) - P_C(y)\|^2), \quad \text{law of cosines} \\ &\iff \|P_C(x) - P_C(y)\|^2 \leq \|x - y\|^2 - \|(x - P_C(x)) - (y - P_C(y))\|^2. \end{aligned}$$

Thus  $P_C$  is firmly nonexpansive. Finally, by Lemma 3, we have

$$\begin{aligned} \|(2P_C(x) - x) - (2P_C(y) - y)\|^2 &= \|2(P_C(x) - P_C(y)) + (1 - 2)(x - y)\|^2, \\ &= 2\|P_C(x) - P_C(y)\|^2 + (1 - 2)\|x - y\|^2 - 2(1 - 2)\|(P_C(x) - x) - (P_C(y) - y)\|^2, \\ &\leq 2[\|x - y\|^2 - \|(x - P_C(x)) - (y - P_C(y))\|^2] - \|x - y\|^2 + 2\|(x - P_C(x)) - (y - P_C(y))\|^2, \\ &= \|x - y\|^2. \end{aligned}$$

□

**Corollary 5** Let  $C_1, C_2 \subseteq \mathbb{R}^m$  be closed nonempty convex sets. Then  $N = \frac{3}{2}P_{C_2}P_{C_1} - \frac{1}{2}I$  is nonexpansive.

**Proof:** Recall that  $\|\cdot\|^2$  is a convex function, so

$$\left\| \frac{1}{2}(x + y) \right\|^2 \leq \frac{1}{2}\|x\|^2 + \frac{1}{2}\|y\|^2, \quad \text{i.e.} \quad \frac{1}{2}\|x + y\|^2 \leq \|x\|^2 + \|y\|^2.$$

Now let  $x, y \in \mathbb{R}^n$ .

$$\begin{aligned} &\frac{1}{2}\|(I - P_{C_2}P_{C_1})(x) - (I - P_{C_2}P_{C_1})(y)\|^2 \\ &= \frac{1}{2}\|(I - P_{C_1})(x) - (I - P_{C_1})(y) + (P_{C_1} - P_{C_2}P_{C_1})(x) - (P_{C_1} - P_{C_2}P_{C_1})(y)\|^2, \\ &\leq \|(I - P_{C_1})(x) - (I - P_{C_1})(y)\|^2 + \|(P_{C_1} - P_{C_2}P_{C_1})(x) - (P_{C_1} - P_{C_2}P_{C_1})(y)\|^2, \\ &\leq \|x - y\|^2 - \|P_{C_1}(x) - P_{C_1}(y)\|^2 + \|P_{C_1}(x) - P_{C_1}(y)\|^2 - \|P_{C_2}P_{C_1}(x) - P_{C_2}P_{C_1}(y)\|^2, \end{aligned}$$

where we apply Lemma 4 twice to get the last inequality. Thus,

$$\|P_{C_2}P_{C_1}(x) - P_{C_2}P_{C_1}(y)\|^2 + \frac{1}{2}\|(I - P_{C_2}P_{C_1})(x) - (I - P_{C_2}P_{C_1})(y)\|^2 \leq \|x - y\|^2.$$

Therefore, by Lemma 3, we have

$$\begin{aligned} \|N(x) - N(y)\|^2 &= \left\| \frac{3}{2}(P_{C_2}P_{C_1}(x) - P_{C_2}P_{C_1}(y)) - \frac{1}{2}(x - y) \right\|^2, \\ &= \frac{3}{2}\|P_{C_2}P_{C_1}(x) - P_{C_2}P_{C_1}(y)\|^2 - \frac{1}{2}\|x - y\|^2 + \frac{3}{4}\|(I - P_{C_2}P_{C_1})(x) - (I - P_{C_2}P_{C_1})(y)\|^2, \\ &\leq \frac{1}{2}[3\|x - y\|^2 - \|x - y\|^3], \\ &= \|x - y\|^2. \end{aligned}$$

□

We'll use the following simple fact.

**Exercise 1**  $\text{Fix}(P_C) = C$ .

**Proposition 6** Let  $C_1, C_2 \subseteq \mathbb{R}^n$  be closed, convex sets such that  $C_1 \cap C_2 \neq \emptyset$ . Then

$$C_1 \cap C_2 = \text{Fix}(P_{C_2} \circ P_{C_1}) = \text{Fix}\left(\frac{3}{2}P_{C_2} \circ P_{C_1} - \frac{1}{2}I\right).$$

**Proof:**  $N = P_{C_2} \circ P_{C_1}$  is nonexpansive by Lemma 4, and the fact that the compositions of nonexpansive maps are nonexpansive. Now, let  $x \in C_1 \cap C_2$ . Then  $P_{C_1}(x) = x$  and  $P_{C_2}(x) = x$ . Thus,  $(P_{C_2} \circ P_{C_1})(x) = x$  and  $x \in \text{Fix}(N)$ .

Now suppose,  $x \in \text{Fix}(N)$ . Then

$$x = P_{C_2} \circ P_{C_1}(x),$$

and so  $x \in C_2$ . We consider three cases:

1. Suppose  $P_{C_1}(x) \in C_2$ . Then  $x = P_{C_2}P_{C_1}x = P_{C_1}x$ , so  $x \in C_1 \cap C_2$ .
2. Suppose  $x \in C_1$ . Then  $x \in C_1 \cap C_2$ .
3. Suppose  $x \notin C_1$  and  $P_{C_1}x \notin C_2$ . Then  $\forall y \in C_1 \cap C_2$ , we have

$$\begin{aligned} \|x - y\| &= \|P_{C_2}P_{C_1}(x) - P_{C_2}P_{C_1}(y)\|, \\ &< \|P_{C_1}(x) - P_{C_1}(y)\|, \quad (P_{C_1}(y) = y \in C_2 \quad \text{and} \quad P_{C_1}(x) \notin C_2, \\ &< \|x - y\|, \quad (x \notin C_1 \quad \text{and} \quad y \in C_1 \cap C_2). \end{aligned}$$

This is a contradiction! So  $x \in C_1 \cap C_2$ .

The equality  $\text{Fix}(P_{C_2}P_{C_1}) = \text{Fix}(\frac{3}{2}P_{C_2}P_{C_1} - \frac{1}{2}I)$  follows because  $P_{C_2}P_{C_1} = \frac{2}{3}(\frac{3}{2}P_{C_2}P_{C_1} + (1 - \frac{3}{2})I) + \frac{1}{3}I$ .

□

**Theorem 7** Suppose  $C_1, C_2 \subseteq \mathbb{R}^n$  are closed convex sets such that  $C_1 \cap C_2 \neq \emptyset$ . Let  $z^0 \in \mathbb{R}^n$ . Then the Method of Alternating Projections

$$z^{k+1} = P_{C_2} P_{C_1} z^k$$

converges to an element of  $C_1 \cap C_2$ .

**Proof:** Let  $N = \frac{3}{2}P_{C_2}P_{C_1} - \frac{1}{2}I$ , apply KM iteration theorem with  $\lambda = \frac{2}{3}$  and observe that

$$N_\lambda = (1 - \lambda)I + \lambda N = \frac{1}{3}I + P_{C_2}P_{C_1} - \frac{1}{3}I = P_{C_2}P_{C_1}.$$

□

**Remark 1** 1. In general, the method of alternating projections can converge arbitrarily slowly!

2. If  $C_1 \cap C_2 = \emptyset$ , then under certain conditions

$$\|z^k - P_{C_1}(z^k)\| \rightarrow \inf_{z \in C_2, w \in C_1} \|z - w\|$$

and  $z^k - P_{C_1}(z^k)$  converges to the gap vector  $v = z^* - w^*$ , where  $(z^*, w^*) \in \operatorname{argmin}_{z \in C_2, w \in C_1} \|z - w\|$

3. Was originally introduced by van-Neumann and Halperin in the 1930s.

Returning to the LP feasibility problem, i.e.: we let  $C_1$  be the solutions to

$$\begin{bmatrix} A & 0 & 0 \\ 0 & A^T & 1 \\ c^T & -b^T & 0 \end{bmatrix} \begin{bmatrix} x^* \\ y^* \\ s^* \end{bmatrix} = \begin{bmatrix} b \\ c \\ 0 \end{bmatrix}$$

and the set  $C_2 = \{(x, y, s) \in \mathbb{R}^{m+2n} \mid x, s \geq 0\}$ . Then we consider the feasibility problem:

$$\begin{bmatrix} x^* \\ y^* \\ s^* \end{bmatrix} \in C_1 \cap C_2.$$

Let's apply the MAP algorithm. Must compute projections first. Let  $z = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$ .

- The projection onto  $C_2$  is a simple thresholding operation:

$$P_{C_2}(z) = \begin{bmatrix} \max\{x, 0\} \\ y \\ \max\{s, 0\} \end{bmatrix}.$$

- Computing  $P_{C_1}(z)$  requires a linear system solve. Let

$$D = \begin{bmatrix} A & 0 & 0 \\ 0 & A^T & 1 \\ c^T & -b^T & 0 \end{bmatrix}$$

then

$$P_{C_1}(z) = z - D^\dagger \left( Dz - \begin{bmatrix} b \\ c \end{bmatrix} \right),$$

where  $D^\dagger$  is the Moore-Penrose inverse. When  $D$  has full rank,

$$P_{C_1}(z) = z - D^\dagger \left( DD^\dagger \right)^{-1} \left( Dz - \begin{bmatrix} b \\ c \end{bmatrix} \right).$$

- The matrix  $D^\dagger$  can be computed offline or one can solve the equation at each iteration. If one intends to run the algorithm for a long time, it may be a good idea to precompute  $D^\dagger$ .
- Furthermore, it can be shown that, given  $z^0 = \begin{bmatrix} x^0 \\ y^0 \\ s^0 \end{bmatrix}$ ,

$$z^{k+1} = P_{C_2} P_{C_1}(z^k),$$

the MAP sequence converges *linearly*.

**Theorem 8** *There exists  $\delta \in (0, 1)$  such that for all  $k \in \mathbb{N}$*

$$\text{dist}_{C_1 \cap C_2}(z^{k+1}) \leq \delta \text{dist}_{C_1 \cap C_2}(z^k).$$

*Hence, for all  $k \in \mathbb{N}$   $\text{dist}_{C_1 \cap C_2}(z^k) \leq \delta^k \text{dist}_{C_1 \cap C_2}(z^0)$ .*

In general,  $\delta$  depends on the “angle” between  $C_1$  and  $C_2$ . The more transversely they meet, the better.