

# ECE 586: Vector Space Methods

## Lecture 22: Alternating Projection

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# Alternating Projection for Subspaces

Let  $P_U$  and  $P_W$  be orthogonal projections onto closed subspaces  $U$  and  $W$  of a Hilbert space  $V$ . For an arbitrary  $\underline{v}_0 \in V$ , what is the behavior of the vector sequence  $\underline{v}_n$  generated by **alternating projection**:

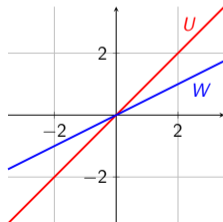
$$\underline{v}_{n+1} = \begin{cases} P_U \underline{v}_n & \text{if } n \text{ is even} \\ P_W \underline{v}_n & \text{if } n \text{ is odd.} \end{cases}$$

Since  $P_U \underline{v} = \underline{v}$  (resp.  $P_W \underline{v} = \underline{v}$ ) if and only if  $\underline{v} \in U$  (resp.  $\underline{v} \in W$ ), it is easy to see that any vector  $\underline{v} \in U \cap W$  is a fixed point of this recursion.

Letting  $P_{U \cap W}$  denote the orthogonal projection onto  $U \cap W$ , one can show that the **sequence  $\underline{v}_n$  converges to  $P_{U \cap W} \underline{v}_0$** .

## Theorem

*The sequence  $\underline{v}_n$  converges to  $P_{U \cap W} \underline{v}_0$ , its projection onto  $U \cap W$ .*



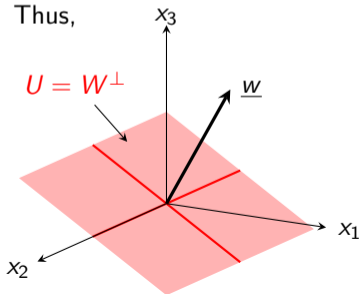
## 4.6: Projection onto Hyperplane Subspaces

The orthogonal projection of  $\underline{v} \in V$  onto a 1D subspace  $W = \text{span}(\underline{w})$  is

$$P_W(\underline{v}) = \frac{\langle \underline{v}, \underline{w} \rangle}{\|\underline{w}\|^2} \underline{w}.$$

A subset of  $V$  that satisfies a single linear equality of the form  $\langle \underline{v}, \underline{w} \rangle = 0$  is a subspace  $U \subset V$  with co-dimension one (i.e.,  $\dim(U) = \dim(V) - 1$ ). Also,  $U = W^\perp$  for a 1D subspace  $W$  and  $U$  is a hyperplane containing  $\underline{0}$ . Thus,

$$P_U(\underline{v}) = P_{W^\perp}(\underline{v}) = \underline{v} - \frac{\langle \underline{v}, \underline{w} \rangle}{\|\underline{w}\|^2} \underline{w}.$$



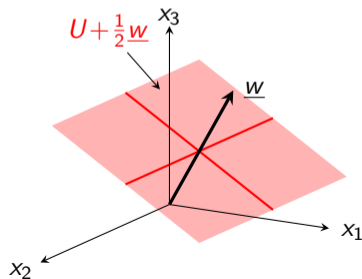
## 4.6: Projection onto Hyperplanes

The linear equation  $\langle \underline{v}, \underline{w} \rangle = c$  defines a shifted subspace  $U + \underline{v}_0$  (for any  $\underline{v}_0 \in V$  satisfying  $\langle \underline{v}_0, \underline{w} \rangle = c$ ) with co-dimension one (i.e., a hyperplane):

$$\langle \underline{v}, \underline{w} \rangle = \langle \underline{u} + \underline{v}_0, \underline{w} \rangle = \langle \underline{u}, \underline{w} \rangle + \langle \underline{v}_0, \underline{w} \rangle = 0 + c = c.$$

One can project onto  $U + \underline{v}_0$  by shifting, projecting, and shifting back:

$$\begin{aligned} P_{U+\underline{v}_0}(\underline{v}) &= \left( (\underline{v} - \underline{v}_0) - \frac{\langle \underline{v} - \underline{v}_0, \underline{w} \rangle}{\|\underline{w}\|^2} \underline{w} \right) + \underline{v}_0 \\ &= \underline{v} - \frac{\langle \underline{v}, \underline{w} \rangle - c}{\|\underline{w}\|^2} \underline{w}, \end{aligned}$$



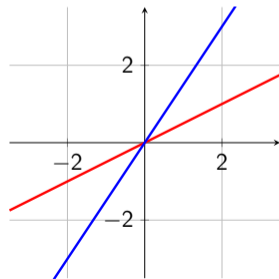
## 4.6: Solving Linear Equations via Alternating Projection

Let  $A \in \mathbb{R}^{m \times n}$  and  $\underline{b} \in \mathbb{R}^m$  be define a set of  $m$  linear equations in  $n$  variables with at least one solution.

The goal is to use alternating projection find a solution  $\underline{x}^*$  such that  $A\underline{x}^* = \underline{b}$ . If  $\underline{b} = \underline{0}$ , then the set of solutions is a subspace equal to the null space of  $A$ ,

$$\mathcal{N}(A) = \{\underline{x} \in \mathbb{R}^n \mid A\underline{x} = \underline{0}\} = \bigcap_{i=1}^m \left\{ \underline{x} \in \mathbb{R}^n \mid \sum_{j=0}^n a_{i,j}x_j = b_i = 0 \right\}.$$

The result follows because  $\mathcal{N}(A)$  is the intersection of  $m$  hyperplane subspaces (i.e., subspaces of dimension  $n - 1$ ). But, what if  $\underline{b} \neq \underline{0}$ ?



## 4.6: Kaczmarz's Algorithm

The idea is to **iteratively project a candidate vector onto linear equality constraints**. For a matrix  $A \in \mathbb{R}^{m \times n}$  and vector  $\underline{b} \in \mathbb{R}^m$ , the algorithm starts from  $\underline{v}_0 = \underline{0}$  and defines  $\underline{v}_{i+1}$  to be the projection of  $\underline{v}_i$  onto the set

$$W_i = \left\{ \underline{v} \in \mathbb{R}^n \mid \sum_{k=1}^n a_{\sigma(i),k} v_k = b_{\sigma(i)} \right\},$$

where  $\sigma(i) = (i \bmod m) + 1$ .

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Using the previously derived projection formula, this gives

$$\underline{v}_{i+1} = (1 - s)\underline{v}_i + s P_{W_{\sigma(i)}}(\underline{v}_i) = \underline{v}_i - s \frac{\langle \underline{v}_i, \underline{a}_{\sigma(i)} \rangle - b_{\sigma(i)}}{\|\underline{a}_{\sigma(i)}\|^2} \underline{a}_{\sigma(i)},$$

where  $s \in (0, 1]$  is the **step-size** and  $\underline{a}_j$  is the  $j$ -th row of the matrix  $A$ .

Note: The true projection uses  $s = 1$  but  $s < 1$  may work better if  $\underline{b} \notin \mathcal{R}(A)$ .

## 4.6: Alternating Projection onto Convex Sets

Let  $C_1, C_2, \dots, C_m$  be closed convex subsets in a Hilbert space  $V$ . The **alternating projection algorithm** finds a point in their intersection. Starting from any  $\underline{v}_0 \in V$ , the alternating projection algorithm computes

$$\underline{v}_{i+1} = (1 - s)\underline{v}_i + s P_{C_{\sigma(i)}}(\underline{v}_i),$$

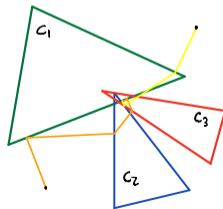
where  $\sigma(i) = (i \bmod m) + 1$  and  $s \in (0, 1]$ .

### Remark

The intersection of convex sets is convex, so  $C \triangleq \bigcap_{i=1}^m C_i$  is convex set. Ideally, alternating projection would give  $\underline{v}_i \rightarrow P_C(\underline{v}_0)$  but it does not :-)

### Theorem (Bregman)

For finite-dimensional  $V$ , there is some  $\underline{v} \in \bigcap_{i=1}^m C_i$  such that  $\underline{v}_i \rightarrow \underline{v}$ .



## 4.6: Orthogonal Projection onto Half Spaces

For  $\underline{w} \in V$ , let  $H = \{\underline{v} \in V \mid \langle \underline{v}, \underline{w} \rangle \geq c\}$  be a closed convex half space.

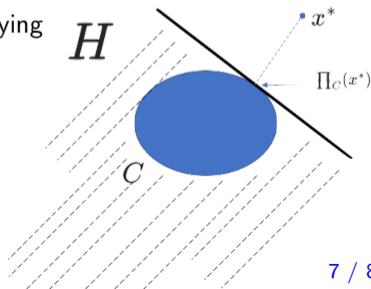
- For  $\underline{v} \in H$ , the projection satisfies  $P_H(\underline{v}) = \underline{v}$
- For  $\underline{v} \notin H$ , the projection satisfies  $P_H(\underline{v}) = P_{U+\underline{v}_0}(\underline{v})$  because the closest point in  $H$  achieves the inequality with equality

It follows that

$$P_H(\underline{v}) = \begin{cases} \underline{v} & \text{if } \langle \underline{v}, \underline{w} \rangle \geq c \\ \underline{v} - \frac{\langle \underline{v}, \underline{w} \rangle - c}{\|\underline{w}\|^2} \underline{w} & \text{if } \langle \underline{v}, \underline{w} \rangle < c \end{cases}$$

Thus, alternating projection can find a feasible vector  $\underline{x} \in \mathbb{R}^3$  satisfying

$$\begin{aligned} 2x_1 - x_2 + x_3 &\geq -1 \\ x_1 + 2x_3 &\geq 2 \\ -7x_1 + 4x_2 - 6x_3 &\geq 1 \\ -3x_1 + x_2 - 2x_3 &\geq 0 \end{aligned}$$





- To continue studying after this video –
  - Try the required reading: Course Notes EF 4.6
  - Look at the Mini-Project Handout on Alternating Projection
  - Also, look at the related problems in Assignments 8 and 9