

## 2.2

### 2.2.1

We are trying to prove that:  $Q$  is convex, then  $\mathbb{R}_+(Q)$  is convex too.

Consider 2 elements in the form of  $ax, by$  taken out from the set  $\mathbb{R}_+(Q)$ :

$$\begin{aligned} ax \in \mathbb{R}_+(Q), by \in \mathbb{R}_+(Q); x, y \in Q, a, b \geq 0 \\ rx + (1-r)y \in Q \quad \forall r \in (0, 1) \\ \implies \forall u \geq 0 : urx + (1-r)uy \in \mathbb{R}_+(Q) \end{aligned} \tag{2.2.1.1}$$

Now let's consider a substitution of expressions:

$$\begin{aligned} ur = \lambda a \quad \lambda \in (0, 1), a \geq 0 \\ (1-r)u = (1-\lambda)by \quad b \geq 0 \\ \implies u - ur = (1-\lambda)by \\ \implies u - \lambda a = (1-\lambda)by \\ u = (1-\lambda)by + \lambda a \geq 0 \\ r = \frac{\lambda a}{u} = \frac{\lambda a}{(1-\lambda)by + \lambda a} \in (0, 1) \end{aligned} \tag{2.2.1.2}$$

By the choice of  $\lambda \in (0, 1), a, b \geq 0$ , we preserve the property of  $u \geq 0$  and  $r \in (0, 1)$ . Consider again the expression:

$$\begin{aligned} \forall u \geq 0, r \in (0, 1) : urx + (1-r)uy \in \mathbb{R}_+(Q) \\ \implies \lambda ax + (1-\lambda)by \in \mathbb{R}_+(Q); \lambda \in (0, 1), a, b \geq 0 \end{aligned} \tag{2.2.1.3}$$

sub with 2.2.1.2

Using the fact that  $ax, by \in Q$ , the last expression is a convex combinations of the 2 points, and it's presented in the set  $\mathbb{R}_+$ , therefore, the set  $\mathbb{R}_+(Q)$  is convex.

### 2.2.2

We wish to prove that if  $Q_1, Q_2$  is convex, then the set  $Q_1 + Q_2$  is also a convex set. From the definition of set addition we have:

$$Q_1 + Q_2 = \{q_1 + q_2 : q_1 \in Q_1, q_2 \in Q_2\}$$

Then considering choosing  $x, y \in (Q_1 + Q_2)$  and using the definition we can characterize  $x, y$  as:

$$\begin{aligned} \exists q_1 \in Q_1, q_2 \in Q_2 : x = q_1 + q_2 \\ \exists q_3 \in Q_1, q_4 \in Q_2 : y = q_3 + q_4 \end{aligned} \tag{2.2.2.1}$$

Consider the convex combinations of the 2 points  $x, y$  we have:

$$\begin{aligned} \lambda x = \lambda(q_1 + q_2) \\ (1-\lambda)y = (1-\lambda)(q_3 + q_4) \\ \implies \lambda x + (1-\lambda)y = \lambda q_1 + (1-\lambda)q_3 + \lambda q_2 + (1-\lambda)q_4 \end{aligned} \tag{2.2.2.2}$$

And notice that  $\lambda q_1 + (1-\lambda)q_3 \in Q_1$  by convexity of  $Q_1$ , by a similar token the element  $\lambda q_2 + (1-\lambda)q_4$  is also in the set  $Q_2$  as well, therefore the convex combination of  $x, y$  is in the set  $(Q_1 + Q_2)$ .

### 2.2.3

We wish to prove that the intersections of convex set is still a convex set.

Set the intersection of convex sets be  $\bigcap_{i \in I} Q_i$ , where the set  $I$  is an indexing set, then we consider choosing  $x, y \in \bigcap_{i \in I} Q_i$ , which means that  $\forall i \in I : x, y \in Q_i$ . Then:

$$\begin{aligned} \lambda x + (1 - \lambda)y &\in Q_i \forall i \in I \\ \lambda x + (1 - \lambda)y &\in \bigcap_{i \in I} Q_i \end{aligned} \tag{2.2.3.1}$$

□

### 2.2.4

We wish to prove that if  $Q \subset \mathbf{E}$  is convex,  $L \subset \mathbf{Y}$  is convex,  $A : \mathbf{E} \mapsto Y$ , then  $A(Q), A^{-1}(L)$  are convex, where  $A^{-1}$  is the pre-image of the linear operator.

Choose  $x, y$  from the image of  $A$ :  $x, y \in A(Q)$ , then  $\exists u \in Q : A(u) = x, \exists v \in Q : A(v) = y$ , by the definition of an image of the operator  $A$ . Consider the convex combinations of  $x, y$ , we have:

$$\begin{aligned} \lambda x + (1 - \lambda)y &= \lambda A(u) + (1 - \lambda)A(v) \quad \forall \lambda \in (0, 1) \\ &= A(\lambda u + (1 - \lambda)v) \end{aligned} \tag{2.2.4.1}$$

using the fact that  $Q$  is a convex set,  $\lambda u + (1 - \lambda)v$  is in  $Q$ , and  $\lambda x + (1 - \lambda)y$  is in the range of the operator  $A$ .

Consider choices of  $x, y$  from the pre-image of  $A$  for  $x, y$ , let  $U := \{u \in L : A^{-1}(u) = x\}$  and  $V := \{v \in L : A^{-1}(v) = y\}$ . Then consider the convex combinations of  $x, y$ :

$$\begin{aligned} \lambda x + (1 - \lambda)y &= \lambda A^{-1}(U) + (1 - \lambda)A^{-1}(V) \quad \forall \lambda \in (0, 1) \\ &= A^{-1}(\lambda U + (1 - \lambda)V) \end{aligned} \tag{2.2.4.2}$$

By the convexity of  $L$ , the set  $\lambda U + (1 - \lambda)V$  is a subset of  $Q$ , therefore, the pre-image of the convex combinations of  $x, y$  is still a preimage of  $A$ , therefore the set of pre-images of  $A$  is convex if  $L$  is convex.

## 2.6

Let  $Q \subseteq \mathbf{E}$ ,  $Q$  convex and  $k \in \mathbb{N}$ , then the convex combinations of  $k$  points in the set  $Q$  is still in  $Q$ .

Inductively we assume that the convex combinations of  $k - 1$  points from  $Q$  is in the set  $Q$ , and we wish to show that the convex combinations of  $k$  point will be in  $Q$ , let  $k \geq 2$ . Define:

$$\begin{aligned} S_{k-1} &:= \left\{ \sum_{i=1}^{k-1} \lambda_i x_i : \lambda \in \Delta_{k-1}, x_i \in Q \forall 1 \leq i \leq k-1 \right\} \\ S_{k-1} &\subseteq Q \quad \text{Inductive Hypothesis} \end{aligned} \tag{2.6.1}$$

Then we consider the convex combinations of  $k$  points, which is any instance of the set  $S_k$

$$\begin{aligned}
\sum_{i=1}^k \lambda_i x_i &= \left( \sum_{i=1}^{k-1} \lambda_i x_i \right) + \lambda_k x_k \\
&= (1 - \lambda_k) \left( \sum_{i=1}^{k-1} \frac{\lambda_i x_i}{1 - \lambda_k} \right) + \lambda_k x_k \\
\lambda \in \Delta_k &\implies \sum_{i=1}^{k-1} \frac{\lambda_i}{1 - \lambda_k} = 1 \\
\lambda \in \Delta_k &\implies 0 \leq \lambda_k \leq 1
\end{aligned} \tag{2.6.2}$$

We had expressed the convex combinations of  $k$  points into the convex combinations of 2 points, where one of them is from the set  $S_{k-1}$ . Using the fact that  $S_k$  is a subset of  $Q$ , both points are in  $Q$ , using the definition convex set, the convex combinations of these 2 points are in the set  $Q$  as well. Therefore, for any instance of  $S_k$ ,  $S_k \subseteq Q$ .

The base case is  $S_2$ , if  $Q$  is convex, then  $S_2 \subseteq Q$  by the definition of a convex set. Therefore, by the principle of mathematics induction,  $\forall k \in \mathbb{N} : S_k \subseteq Q$ .

## 2.8

We wish to prove that:

$$\text{conv}(Q) = T; \quad T := \left\{ \sum_{i=1}^k \lambda_i x_i : k \in \mathbb{N}, \{x_i\}_1^k \subseteq Q, \lambda \in \Delta_k \right\}$$

The convex hull of any set  $Q$  is the convex combinations of all the points in the set  $Q$ . The convex cone is:

$$\text{conv}(Q) = \bigcap \{C : C \text{ is convex and } Q \subset C\}$$

It's the intersections of all convex sets that contains  $Q$ .

$Q \in \text{conv}(Q)$  by definition of  $\text{conv}(Q)$ :

$$\begin{aligned}
\forall C, C \text{ convex} : Q \subseteq C, T \subseteq Q \subseteq C &\quad \text{by 2.8} \\
\implies T \subseteq \text{conv}(Q)
\end{aligned} \tag{2.8.1}$$

If  $T$  is convex, then  $\text{conv}(Q)$  is a subset of  $T$ , because  $T$  is one of those convex  $C$  that contains  $Q$ .  $T$  contains  $Q$  because any points in  $Q$  can be added into the convex combinations of  $Q$  by incrementing the value  $k$ . We now wish to show that  $T$  is convex so that  $\text{conv}(Q) \subseteq T$ .

$$\begin{aligned}
u, v &\in T \\
\implies u &\in \sum_{i=1}^{k_1} \beta_i x_i, \quad v \in \sum_{i=1}^{k_2} \alpha_i y_i \\
\implies \lambda u + (1 - \lambda)v &= \lambda \sum_{i=1}^{k_1} \beta_i x_i + (1 - \lambda) \sum_{i=1}^{k_2} \alpha_i y_i \quad \forall \lambda \in (0, 1)
\end{aligned} \tag{2.8.2}$$

Notice that:

$$\begin{aligned}
& \lambda \sum_{i=1}^{k_1} \beta_i + (1-\lambda) \sum_{i=1}^{k_2} \alpha_i = \lambda(1) + (1-\lambda)(1) = 1 \\
& \implies \forall \lambda \in (0, 1) : \begin{bmatrix} \lambda\beta \\ (1-\lambda)\alpha \end{bmatrix} \in \Delta_{k_1+k_2} \\
& \implies \lambda u + (1-\lambda)v \in T \\
& \implies T \text{ is convex}
\end{aligned} \tag{2.8.3}$$

Therefore  $\text{conv}(Q) \subseteq T$ , therefore  $T = \text{conv}(Q)$ .

## 2.16

We wish to prove that the  $\text{dist}_Q$  is a 1-Lipschitz continuous function:

$$|\text{dist}_Q(x) - \text{dist}_Q(y)| \leq \|x - y\|_2$$

Choose any  $x, y \in \mathbf{E}$ ,  $z \in Q$ . Then we have:

$$\begin{aligned}
\text{dist}_Q(x) & \leq \|x - z\|_2 \leq \|x - y\|_2 + \|y - z\|_2 \\
\text{dist}_Q(x) & \leq \|x - y\|_2 + \|y - z\|_2
\end{aligned} \tag{2.16.1}$$

This is triangular inequality. Consider for  $y$ :

$$\begin{aligned}
\text{dist}_Q(y) & \leq \|y - z\|_2 \leq \|y - x\|_2 + \|x - z\|_2 \\
\text{dist}_Q(y) & \leq \|y - x\|_2 + \|x - z\|_2
\end{aligned} \tag{2.16.2}$$

Take the difference between the 2 expression, their absolute value is bounded:

$$\begin{aligned}
|\text{dist}_Q(x) - \text{dist}_Q(y)| & \leq \|y - z\|_2 - \|x - z\|_2 \\
|\text{dist}_Q(x) - \text{dist}_Q(y)| & \leq \|x - y\|_2
\end{aligned} \tag{2.16.3}$$

because  $\|y - z\| \leq \|x - z\| + \|x - y\|$ . And the last inequality is imposed using triangular inequality.

## 2.18

We wish to prove that the projection function  $\text{proj}_Q(x)$  is a 1-Lipschitz function when the set  $Q$  is closed and convex.

We wish to prove this claim:

$$\|\text{proj}_Q(x_1) - \text{proj}_Q(x_2)\|_2^2 \leq \left\langle \text{proj}_Q(x_1) - \text{proj}_Q(x_2), x_1 - x_2 \right\rangle$$

If this is true, then we can use the Cuachy Schwartz inequality, which gives us the 1-Lipschitz continuity:

$$\begin{aligned}
\left\langle \text{proj}_Q(x_1) - \text{proj}_Q(x_2), x_1 - x_2 \right\rangle & \leq \|\text{proj}_Q(x_1) - \text{proj}_Q(x_2)\| \|x_1 - x_2\| \\
\implies \|\text{proj}_Q(x_1) - \text{proj}_Q(x_2)\|_2 & \leq \|x_1 - x_2\|_2
\end{aligned} \tag{2.18.1}$$

Let's prove the claim using the obtuse angle theorem for the projections of any convex set  $Q$ , we have:

$$\begin{aligned}
& \forall x_1, x_2 \in Q : \\
& \left\langle \text{proj}_Q(x_2) - \text{proj}_Q(x_1), x_1 - \text{proj}_Q(x_1) \right\rangle \leq 0 \\
& \left\langle \text{proj}_Q(x_1) - \text{proj}_Q(x_2), x_2 - \text{proj}_Q(x_1) \right\rangle \leq 0
\end{aligned} \tag{2.18.2}$$

Let me add a negative sign so we can combine them:

$$\begin{aligned}
& \forall x_1, x_2 \in Q : \tag{2.18.3} \\
& \left\langle \text{proj}_Q(x_2) - \text{proj}_Q(x_1), x_1 - \text{proj}_Q(x_1) \right\rangle \leq 0 \\
& \left\langle \text{proj}_Q(x_2) - \text{proj}_Q(x_1), \text{proj}_Q(x_1) - x_2 \right\rangle \leq 0 \\
& \xRightarrow{\text{add them}} \left\langle \text{proj}_Q(x_2) - \text{proj}_Q(x_1), x_1 - x_2 + \text{proj}_Q(x_2) - \text{proj}_Q(x_1) \right\rangle \leq 0 \\
& \|\text{proj}_Q(x_2) - \text{proj}_Q(x_1)\|_2^2 - \left\langle \text{proj}_Q(x_2) - \text{proj}_Q(x_1), x_2 - x_1 \right\rangle \leq 0
\end{aligned}$$

The claim is proven, it follows from 2.18.1 that the projection function is 1-Lipschitz continuous.

## 2.23

Let  $K \subseteq \mathbf{E}$  then  $K$  is a convex cone iff the point  $\lambda x + uy$  lies in  $K$  for any 2 points  $x, y \in K$ ,  $u, \lambda \geq 0$ .

We wish to prove it in 2 directions. Consier the cone  $C$ . First, we wish to show that if the set is a convex cone, then  $\lambda x + uy \in C$  for all  $\lambda, u \geq 0$ .

If the set  $C$  is a cone, then:

$$\begin{aligned}
\lambda x &\in C \quad \forall x \in C, \lambda \geq 0 \\
uy &\in C \quad \forall y \in C, u \geq 0
\end{aligned} \tag{2.23.1}$$

Using the fact that  $C$  is also convex, then:

$$\begin{aligned}
r\lambda x + (1-r)uy &\in C \quad \forall r \in (0, 1) \\
t := r\lambda \geq 0 \quad k := (1-r)u &\geq 0 \\
\forall t, k \geq 0 : tx + ky &\in C
\end{aligned} \tag{2.23.2}$$

Done.

Now we wish to prove that if  $\lambda x + uy \in C$  for all points  $x, y \in C$  and  $\lambda, u \geq 0$ , then the set  $C$  is a convex cone.

Suppose that  $x, y \in C$ ;  $\lambda, u \geq 0$ ,  $\lambda x + uy \in C$ , choose  $\lambda = 0, u \geq 0$ , then  $uy \in C \forall u \geq 0$ . And this indicates that the set  $C$  is a cone.

Choose  $\lambda \in (0, 1), u = (1 - \lambda)$ , then both  $\lambda, u \geq 0$  still, but  $\lambda x + (1 - \lambda)y \in C$  which implies that the set  $C$  is convex.

## 2.27

Double Polar Theorem. Given a set  $K$  that is non-empty and a cone, then:

$$(K^\circ)^\circ = (\text{cl} \circ \text{conv})(K)$$

The convex closure of the cone is the same as the polar polar cone.

The set of halfplanes supporting the cone  $K$  is:

$$\mathcal{F}_K = \{a \in \mathbf{E} : \langle x, a \rangle \leq 0 \forall x \in K\} \subset K^\circ \tag{2.27.1}$$

We can construct the closure of the convex hull by intersecting all the supporting halfplanes for the set  $K$ :

$$\begin{aligned}
\text{cl} \circ \text{conv}(K) &= \bigcap_{a \in \mathcal{F}_K} \{x \in \mathbf{E} : \langle a, x \rangle \leq 0\} \\
&= \bigcap_{a \in K^\circ} \{x \in \mathbf{E} : \langle a, x \rangle \leq 0\} \\
&= \{x \in \mathbf{E} : \langle a, x \rangle \leq 0, a \in K^\circ\} \\
&= (K^\circ)^\circ
\end{aligned} \tag{2.27.2}$$

## 2.33

In this section we wish to prove that,  $\forall Q \in \mathbf{E}$ , where  $Q$  is a convex set; and a point  $\bar{x} \in Q$ :

$$T_Q(\bar{x}) = \text{cl}\mathbb{R}_+(Q - \bar{x})$$

The tangent cone generated at the point  $\bar{x}$  is the closure of the cone generated by offsetting the set  $Q$  by  $\bar{x}$ . Firstly if  $\bar{x} \in \text{int}(Q)$ , then the tangent cone is  $\mathbf{E}$ , and  $Q - \bar{x}$  intersect with an  $\epsilon$  around the origin, making  $\text{cl}\mathbb{R}_+(Q - \bar{x}) = \mathbf{E}$  as well.

Now choose a point  $\bar{x} \in \text{cl}(Q) \setminus \text{int}(Q)$ , from the boundary of the set  $Q$ , then we want to show that  $T_Q(\bar{x}) \supseteq \text{cl}\mathbb{R}_+(Q - \bar{x})$ .

$$\begin{aligned}
&\bar{x} \in Q && \text{By def} && (2.33.1) \\
&\forall x \in Q && \text{Our choice} \\
&\frac{1}{n}x \left(1 - \frac{1}{n}\right) \bar{x} \in Q && \text{By } Q \text{ Convex} \\
&\forall u_i : \lim_{i \rightarrow \infty} u_i = 0 \wedge u_i \in [0, 1] \\
&r_n := u_n x + (1 - u_n) \bar{x} && \text{We defined it} \\
&\lim_{n \rightarrow \infty} r_n = \bar{x}, r_n \in Q \\
&\tau_n = \frac{u_n}{\lambda}, \lambda \geq 0 \implies \tau_n \searrow 0 && \text{Our choice} \\
&\lim_{n \rightarrow \infty} \tau_n^{-1}(r_n - \bar{x}) \in T_Q(\bar{x}) && \text{Def of } T_Q(\bar{x}) \\
&\lim_{n \rightarrow \infty} \tau_n^{-1}(r_n - \bar{x}) \\
&= \lim_{n \rightarrow \infty} \frac{\lambda}{u_n} (u_n x + (1 - u_n) \bar{x} - \bar{x}) \\
&= \lim_{n \rightarrow \infty} (\lambda x - \lambda \bar{x}) = \lambda(x - \bar{x}) \in \mathbb{R}_+(Q - \bar{x}) && \text{by Def}
\end{aligned}$$

Note that, the convexity of  $Q$  is important here.  $x_n$  is a sequence of points connecting the between  $x, \bar{x}$ , which are both in the set  $Q$  by definition. Therefore any points that linearly interpolating between then will still be in the set  $Q$  by the convexity of the set  $Q$ .

Next we wish to prove that  $\text{cl}\mathbb{R}_+(Q - \bar{x}) \supseteq T_Q(\bar{x})$ . Fix any  $\bar{x} \in \text{cl}(Q) \setminus \text{int}(Q)$   
Start with the tangent cone at the point  $\bar{x}$ :

$$\begin{aligned}
v \in T_Q(\bar{x}) &\iff v = \lim_{i \rightarrow \infty} \tau_i(x_i - \bar{x}), \tau_i \searrow 0, x_i \in Q \forall i \\
&\lambda_i := \tau_i^{-1} \geq 0 \implies \tau_i^{-1}(x_i - \bar{x}) = \lambda_i(x_i - \bar{x}) \\
x_i \in Q &\implies x_i - \bar{x} \in Q - \bar{x} \quad \forall i \\
&\implies \lambda_i(x_i - \bar{x}) \in \text{cl}\mathbb{R}_+(Q - \bar{x}) \quad \forall i \\
&\implies v \in \text{cl}\mathbb{R}_+(Q - \bar{x})
\end{aligned} \tag{2.33.2}$$

## 2.36

We wish to prove that, for  $Q \subseteq \mathbf{E}$  with  $Q$  being convex and a point  $\bar{x} \in Q$  then the equality:

$$N_Q(\bar{x}) = \{v \in \mathbf{E} : \langle v, x - \bar{x} \rangle \leq 0 \ \forall x \in Q\}$$

Let's start by choosing  $v \in \text{RHS}$  then:

$$\begin{aligned} \langle v, x - \bar{x} \rangle &\leq 0 \ \forall x \in Q & (2.36.1) \\ \text{let: } x_n &\rightarrow \bar{x}, x_n \in Q \ \forall n \\ \langle v, x_n - \bar{x} \rangle &\leq 0 \ \forall x_n \in Q \\ \langle v, x_n - \bar{x} \rangle &\leq 0 \leq o(\|x - \bar{x}\|_2) \ \forall x_n \in Q \\ &\implies v \in N_Q(\bar{x}) \end{aligned}$$

Now consider choosing  $v \in N_Q(\bar{x})$ , consider:

$$\begin{aligned} N_Q(\bar{x}) &= (T_Q(\bar{x}))^\circ && \text{By Lemma 2.35} && (2.36.2) \\ N_Q(\bar{x}) &= \{v \in \mathbf{E} : \langle v, x \rangle \leq 0 \ \forall x \in T_Q(\bar{x})\} && \text{Def of Polar Cone} \\ x \in T_Q(\bar{x}) &= \text{cl}\mathbb{R}_+(Q - \bar{x}) && Q \text{ convex, use 2.33} \\ x &= \lambda(y - \bar{x}), \lambda \geq 0, y \in Q && \text{Def of } \mathbb{R}_+(Q - \bar{x}) \text{ form match} \\ \implies N_Q(\bar{x}) &= \{v \in \mathbf{E} : \lambda \langle v, y - \bar{x} \rangle \leq 0 \ \forall y \in Q\} && \text{Arrived at LHS} \end{aligned}$$

The proof is done.

## 2.37

We wish to prove that the following statements are all equivalent:

- (a)  $v \in N_Q(\bar{x})$
- (b)  $\bar{x} \in \arg \max_{x \in Q} \langle v, x \rangle$
- (c)  $\text{proj}_Q(\bar{x} + \lambda v) = \bar{x} \ \forall \lambda \geq 0$
- (d)  $\text{proj}_Q(\bar{x} + \lambda v) = \bar{x}$  for some  $\bar{\lambda} \geq 0$

We wish to prove (a)  $\implies$  (b)  $\implies$  (c)  $\implies$  (d)  $\implies$  (a) to show the equivalence of the statement.

verify (a)  $\implies$  (b)

$$\begin{aligned} v &\in N_Q(\bar{x}) && (2.37.1) \\ \forall x \in Q : \langle x - \bar{x}, v \rangle &\leq 0 && \text{by 2.36} \\ \forall x \in Q : \langle x, v \rangle &\leq \langle \bar{x}, v \rangle \\ \bar{x} &\in \arg \max_{x \in Q} \langle v, x \rangle \end{aligned}$$

Verify (b)  $\implies$  (c) We wish to say that:

$$\bar{x} \in \arg \max_{x \in Q} \langle v, x \rangle \implies \text{proj}_Q(\bar{x} + \lambda v) = \bar{x} \ \forall \lambda \geq 0$$

From the definition that  $\bar{x}$  is the maximizer for the dot product on  $v$  we have:

$$\begin{aligned} \bar{x} \in \arg \max_{x \in Q} \langle v, x \rangle &\implies \forall x \in Q : \langle x, v \rangle \leq \langle \bar{x}, v \rangle && (2.37.2) \\ \forall x \in Q : \langle x - \bar{x}, v \rangle &\leq 0 \\ \forall x \in Q : \langle x - \bar{x}, \lambda v \rangle &\leq 0 \end{aligned}$$

Take note that the last expression is the Obtuse Angle characterization of projections on to the convex set  $Q$ , consider  $\lambda v = u - \bar{x}$ , then we have:

$$\begin{aligned}
& \forall x \in Q : \langle x - \bar{x}, u - \bar{x} \rangle \leq 0 \quad \forall \lambda \geq 0 \\
& \implies \text{proj}_Q(u) = \bar{x} \\
& = \text{proj}_Q(\bar{x} + \lambda v) \quad \forall \lambda \geq 0
\end{aligned} \tag{2.37.3}$$

Verify (c)  $\implies$  (d):

This is trivial, simply by choose some  $\lambda$  to be  $\bar{\lambda}$  and the statement follows.

Verify (d)  $\implies$  (a)

We wish to show that

$$\exists \bar{\lambda} > 0 : \text{proj}_Q(\bar{x} + \bar{\lambda}v) = \bar{x} \implies v \in N_Q(\bar{x})$$

Consider:

$$\begin{aligned}
& \forall x \in Q : \langle x - \bar{x}, \bar{\lambda}v \rangle \leq 0 & \text{Proj Obtuse Angle} & (2.37.4) \\
& \forall x \in Q : \langle x - \bar{x}, v \rangle \leq 0 & \bar{\lambda} \geq 0 & \\
& \forall x \in Q : \langle x - \bar{x}, v \rangle \leq o(\|x - \bar{x}\|_2) \\
& x_n \rightarrow \bar{x}, x_n \in Q \quad \forall n \\
& \lim_{n \rightarrow \infty} \langle x_n - \bar{x}, v \rangle \leq o(\|x_n - \bar{x}\|_2) \\
& \implies v \in N_Q(\bar{x})
\end{aligned}$$