

# $L^p$ spaces:

Let  $1 \leq p \leq \infty$ . Given a measure space  $(X, \mu)$

$f: X \rightarrow \mathbb{R}/\mathbb{C}$  measurable.

$$\text{Def: } \|f\|_p = \begin{cases} \left( \int |f|^p d\mu \right)^{\frac{1}{p}} & 1 \leq p < \infty \\ \text{esssup } |f| & p = \infty \end{cases}$$

$$\text{Def: } L^p(\mu) = \{f: X \rightarrow \mathbb{R}/\mathbb{C} \text{ mable, } \|f\|_p < \infty\} / \sim$$

$$f \sim g \Leftrightarrow f = g \text{ a.e.}$$

- Vector space
- $\|\cdot\|_p$  a norm.

Prop: Let  $f, g \in L^p(\mu)$ , then  $\|f+g\|_p \leq \|f\|_p + \|g\|_p$

Pf: Consider the unit ball  $B = \{f \in L^p, \|f\|_p \leq 1\}$ .

①  $B$  is convex, let  $f, g \in B$ ,  $\lambda \in (0, 1)$ , we estimate

$$\|(1-t)f + tg\|_p^p = \int ((1-t)f + tg)^p d\mu \leq \int ((1-t)|f| + t|g|)^p d\mu$$

$$\text{Let } h(y) = y^p, \text{ convex.} \implies \int ((1-t)|f| + t|g|)^p d\mu \leq 1$$

$p = \infty$  case similar

②  $\forall f, g$ , if  $\|f\|_p = 0$  or  $\|g\|_p = 0$ , ✓

$$\text{else: } f+g = (\|f\|_p + \|g\|_p) \left( \frac{|f|}{\|f\|_p + \|g\|_p} + \frac{|g|}{\|f\|_p + \|g\|_p} \right)$$

$\underbrace{\quad}_{t} \quad \underbrace{\quad}_{1-t} \quad \underbrace{\quad}_{=m \in B}$

$$\Rightarrow \|f+g\|_p = (\|f\|_p + \|g\|_p) \downarrow^{<1} \quad \checkmark$$

## Hölder's Inequality:

Given  $1 \leq p \leq \infty$ , Define  $p'$  s.t.  $\frac{1}{p} + \frac{1}{p'} = 1$  (dual exponent)

$$\text{Then, } \forall f \in L^p, g \in L^{p'}, |\int fg d\mu| \leq \|f\|_p \|g\|_{p'} \quad \|fg\|_1 \leq (\int |f|^p)^{\frac{1}{p}} (\int |g|^{p'} )^{\frac{1}{p'}}$$

$$\text{Pf: } p=1, \infty : |\int fg d\mu| \leq \int |f||g| d\mu \leq \text{esssup } |f| \int |g| d\mu = \|f\|_{\infty} \|g\|_1,$$

$$f \times \gamma \in (1, \infty)$$

claim:  $\forall a, b > 0, \frac{1}{p} a^p + \frac{1}{p'} b^{p'} \geq ab$ . Young's / Convexity inequality

Consider  $\lambda(x) = bx - \frac{1}{p} x^p$ , concave in  $x$

$$\lambda'(x) = b - x^{p-1} \text{ max when } b^{\frac{1}{p-1}}$$

$$\lambda(b^{\frac{1}{p-1}}) = b^{1+\frac{1}{p-1}} - \frac{1}{p} b^{\frac{p}{p-1}} = (1 - \frac{1}{p}) b^{\frac{p}{p-1}} = \frac{1}{p} b^{p'} \quad \frac{p}{p-1} = p' \Leftrightarrow p = p'(p-1)$$

$$\Rightarrow \forall x, \lambda(x) \leq \frac{1}{p} b^{p'} \Leftrightarrow ba - \frac{1}{p} a^p \leq \frac{1}{p} b^{p'} \quad \Leftarrow p+1=p+p' \quad \Leftarrow \frac{1}{p} + \frac{1}{p'} = 1 \quad \checkmark$$

claim:  $ab = \min_{t \geq 0} \frac{1}{p} t^p a^p + \frac{1}{p'} t^{-p'} b^{p'}$  Sharp Young

① Take  $\lambda(t) = \frac{1}{p} t^p a^p + \frac{1}{p'} t^{-p'} b^{p'}$  convex

$$\lambda(t) = \frac{1}{p} (t^p a^p - t^{-p'} b^{p'}) \text{, vanishes at } (ta)^p = (\frac{b}{t})^{p'}$$

$$\left| \int f g d\mu \right| \stackrel{\text{①}}{\leq} \int \|f\|_p \|g\|_{p'} d\mu \stackrel{\text{sharp}}{\leq} \inf_{t > 0} \left( \frac{1}{p} t^p \|f\|_p^p + \frac{1}{p'} t^{-p'} \|g\|_{p'}^{p'} \right) \quad \int \inf_{t > 0} \left( \frac{1}{p} t^p \|f\|_p^p + \frac{1}{p'} t^{-p'} \|g\|_{p'}^{p'} \right) d\mu \leq \inf_{t > 0} \int \frac{1}{p} t^p \|f\|_p^p$$

$$= \inf_{t > 0} \frac{1}{p} t^p \|f\|_p^p$$

$$\stackrel{\text{sharp}}{=} \|f\|_p \|g\|_{p'}$$

Hölder's inequality suggests the following question:

Given  $f \in L^p$

Define  $L: L^p \rightarrow \mathbb{R}$  by  $L(f) = \int f g d\mu$ .

- Linear
- Continuity

$$\|L\| = \sup_{f \neq 0} \frac{|L(f)|}{\|f\|_p} \leq \sup_{f \neq 0} \frac{\|f\|_p \|g\|_{p'}}{\|f\|_p} = \|g\|_{p'}$$

Given  $f \in L^p$ ,  $\|f\|_p > 0$ , consider  $\tilde{f} := |f|^{p-1} \frac{\overline{f}}{\operatorname{Sign}(f)}$

$$|\tilde{f}|^p = |f|^{p(p-1)} = |f|^p \Rightarrow \int |\tilde{f}|^p = \int |f|^p < \infty$$

$$\begin{aligned} \int f g d\mu &= \int |g|^{p-1} \frac{\operatorname{Sign}(g) f}{|f|} d\mu \\ &= \int |g|^{p-1} |f| \frac{f}{|f|} d\mu = \|g\|_{p'}^{p-1} \|f\|_p \end{aligned}$$

$$\text{Taking in } \frac{L(\tilde{f})}{\|\tilde{f}\|_p^p} = \frac{\int \tilde{f} g d\mu}{\|\tilde{f}\|_p^p} = \frac{\|g\|_{p'}^{p-1} \|f\|_p}{\|g\|_{p'}^{p-1}} = \|g\|_{p'}^{p(1-\frac{1}{p})} = \|g\|_{p'}$$

So  $L^p \hookrightarrow (L^p)^*$  injective.

$$f \mapsto \mathbb{1}_f, \quad \mathbb{1}_f(f) = \int f g d\mu$$

↓  
surjective? Yes  $P < \infty$   
No  $P = \infty$

Equality in ①  $\Leftrightarrow \exists t, \text{ s.t. } fg = e^{it} |fg|$

$$\begin{aligned} \text{Equality in } \int |fg| d\mu &\leq \frac{1}{P} \|f\|_P + \frac{1}{P} \|g\|_P \\ \Leftrightarrow |f|^P &= |g|^P \end{aligned}$$

Equality in ②  $\Leftrightarrow \exists t \text{ constant, } (t|f|)^P = (t|g|)^P \text{ a.e.}$

Minkowski's inequality:

Special case: triangle inequality.

Given  $(X, \mu), (Y, \nu)$  s.t. Fubini holds.  $f: X \times Y \rightarrow \mathbb{R}$  measurable.

$$x \xrightarrow{f(\cdot, y)} f(x, y) \quad \left\| \int_Y f(\cdot, y) d\nu(y) \right\|_{L^p(\mu)} \leq \int_Y \|f(\cdot, y)\|_{L^p(\mu)} d\nu(y) \quad f(\cdot, y) = X \rightarrow f(x, y) = k_y(x)$$

$$\text{That is, } \left( \int_X \left\| f(x, y) \right\|_p^p d\mu(x) \right)^{\frac{1}{p}} \leq \int_Y \left( \int_X |f(x, y)|^p d\mu(x) \right)^{\frac{1}{p}} d\nu(y).$$

$$Pf = f \circ 0 \quad \text{Set } H(x) = \int_Y f(x, y) d\nu(y)$$

$$\begin{aligned} (PH)^P &= \int_X H^P(x) d\mu(x) = \int_X \left( \int_Y f(x, y) d\nu(y) \right)^P d\mu(x) \\ &\stackrel{\text{Fubini}}{=} \int_Y \int_X H^P(x) f(x, y) d\mu(x) d\nu(y) \\ &\stackrel{\text{Holder}}{\leq} \int_Y \|H\|_P^{P-1} \left( \int_X |f(x, y)|^P d\mu(x) \right)^{\frac{1}{P}} d\nu(y). \end{aligned}$$

$$\|H\|_P \leq \int_Y \left( \int_X |f(x, y)|^P d\mu(x) \right)^{\frac{1}{P}} d\nu(y)$$

Recall: Equality in Hölder's ineq

$$P=1, P'=\infty \quad \left\| \int f g \right\| \leq \int |f| |g| \leq \text{essup} |g| \int |f| = \|g\|_\infty \|f\|_1$$

$$\text{equality} = |g(x)| = \text{essup} |g| \text{ a.e. on } \{x \mid f(x) \neq 0\}.$$

Minkowski's ineq:

$$\left( \int_X \int_Y |f(x, y)|^p d\nu(y) d\mu(x) \right)^{\frac{1}{p}} \leq \int_Y \left( \int_X |f(x, y)|^p d\mu(x) \right)^{\frac{1}{p}} d\nu(y)$$

$$\text{Let } Y = \{1, 2\}, \nu(1) = \nu(2) = \frac{1}{2} \Rightarrow \text{triangle ineq}$$

Key step:

$$\|H\|_P^P = \int_X H(x) P^{-1} \int_Y f(x,y) d\nu(y) d\mu(x) \stackrel{Fbn}{=} \int_Y \int_X H(x) P^{-1} f(x,y) d\mu(x) d\nu(y)$$

Equality:  $H(x,y)$  constant multiple of  $f(x,y)|^P$  Apply Holder.

$$|f(x,y)| = \lambda y + (P-1)\frac{P}{P}$$

$$P=1 \text{ case: } \int_X \int_Y f(x,y) d\nu(y) d\mu(x) \stackrel{Fbn}{=} \int_Y \int_X f(x,y) d\mu(x) d\nu(y)$$

So  $f \geq 0 \Rightarrow$  equality always hold.

Thm: Let  $(X, \mu)$  measure space,  $1 \leq P \leq \infty$ , then  $L_P^n$  complete.

Pf: Given a Cauchy seq  $\{x_n\}$

Key: Choose a ("fast" Cauchy sequence) subsequence  $\{x_{n_k}\}$

s.t.  $\|x_{n_k} - x_m\| \leq 2^{-k} \quad \forall m > n_k$ .

Idea: Write  $f_{n_k} = \frac{P}{P} + \sum_{i=1}^k (f_{n_i} - f_{n_{i-1}})$ ,  $\sum \|f_{n_i} - f_{n_{i-1}}\|$  converges.

Consider  $g_k(x) = |f_{n_1}| + \sum_{i=1}^k |f_{n_i} - f_{n_{i-1}}| \in L^P$

So  $g_k \nearrow g$  ptwise by MCT. \|g\_k\|\_P \leq \|f\_{n\_1}\|\_P + \dots + \|f\_{n\_k}\|\_P \Rightarrow \|g\|\_P < \infty

$\|g_k\|_P^P = \int g_k^P d\mu = \int |g_k(x)|^P dx \quad \lim g_k(x) \text{ exists } \forall x \text{ (may be } \infty)$

$\int g^P = \lim \int g_k^P$  by MCT.

?

Hanner's inequality: Let  $1 < P < \infty$

① If  $1 < P \leq 2$ , then

$$\|f+g\|_P^P + \|f-g\|_P^P \geq (\|f\|_P + \|g\|_P)^P + (\|f\|_P - \|g\|_P)^P$$

② If  $2 \leq P < \infty$ ,  $\dots \leq \dots \dots \dots$

In particular,  $\|f+g\|_2^2 + \|f-g\|_2^2 \geq (\|f\|_2^2 + \|g\|_2^2)$

"Parallelogram"

Plug in  $f \rightarrow f+g, g \rightarrow f-g$

$$\Rightarrow 2^P (\|f\|_P^P + \|g\|_P^P) \geq (\|f+g\|_P^P + \|f-g\|_P^P) + |\|f+g\|_P - \|f-g\|_P|^P$$

Pf:  $P=2$ :

$$\begin{aligned}
2\|f+g\|_P^2 + \|f-g\|_P^2 &= \int |f+g|(\bar{f}+\bar{g}) + |f-g|(\bar{f}-\bar{g}) du \\
&= \int_0^1 |\bar{f}|^2 + 2|\bar{g}|^2 \\
&= 2\|f\|_P^2 + \|g\|_P^2 \\
&= (\|f\|_P + \|g\|_P)^2 + (\|f\|_P - \|g\|_P)^2
\end{aligned}$$

$\Rightarrow$  Assume  $\|f\| = 1 > R = \|g\| > 0$

Construct a function  $F: [0, 1] \rightarrow \mathbb{R}$

$$F(r) = \alpha(r) + \beta(r) R^P \quad (\alpha, \beta \text{ TBD}) \quad \min_{0 \leq r \leq 1} F(r) = (1+R)^P + (1-R)^P$$

Moreover:  $\forall A, B \in \mathbb{C}$ , will show that

$$|A+B|^P + |A-B|^P \stackrel{(2)}{\leq} \alpha(r)|A|^P + \beta(r)|B|^P$$

Given such  $F$ , we estimate

$$\begin{aligned}
\|f+g\|_P^P + \|f-g\|_P^P &= \int |f(x)+g(x)|^P + |f(x)-g(x)|^P \\
&\stackrel{(2)}{\leq} \int \alpha(r) |f(x)|^P + \beta(r) |g(x)|^P \quad (\forall r \in [0, 1]) \\
&= \alpha(r) \|f\|_P^P + \beta(r) \|g\|_P^P \\
&= \alpha(r) + \beta(r) R^P \\
&\stackrel{(1)}{\leq} (1+R)^P + (1-R)^P
\end{aligned}$$

Now to construct  $F$

$$\text{Choose } \alpha(r) = (1+r)^{P-1} + (1-r)^{P-1}$$

$$\beta(r) = r^{1-P} ((1+r)^{P-1} - (1-r)^{P-1})$$

$$\begin{aligned}
F'(r) &= \alpha' + \beta' R^P \\
&= (P-1) \left( (1+r)^{P-2} - (1-r)^{P-2} + R^P \left( -r^{-P} ((1+r)^{P-1} + (1-r)^{P-1}) \right) \right. \\
&\quad \left. - r^{1-P} ((1+r)^{P-2} + (1-r)^{P-2}) \right)
\end{aligned}$$

$$\begin{aligned}
\frac{F'}{P-1} &= (1+r)^{P-2} \left( 1 - \left(\frac{R}{r}\right)^P (1+r-r) \right) + (1-r)^{P-2} \left( -1 + \left(\frac{R}{r}\right)^P (1-r+r) \right) \\
&= \left( 1 - \left(\frac{R}{r}\right)^P \right) \left( (1+r)^{P-2} - (1-r)^{P-2} \right)
\end{aligned}$$

$$F'(R) = 0, \quad F'(r) \begin{cases} < 0 & \text{if } 0 < r < R \\ > 0 & \text{if } R < r < 1 \end{cases}$$

$\xrightarrow[R]{\quad} \text{Plug in } r=R \Rightarrow \text{get (1).}$

For ②, NB if  $A, B > 0$ , then can rescale from ①.

$A = |A|e^{i\theta}, B = |B|e^{i\varphi}$ , rotate s.t.  $\varphi = 0$ .

$$\begin{aligned} |A+B|^P + |A-B|^P &= (|A|^2 + |B|^2 + 2\cos\theta|A||B|)^{\frac{P}{2}} + (|A|^2 + |B|^2 - 2\cos\theta|A||B|)^{\frac{P}{2}} \\ &= 2(|A|^2 + |B|^2 + 2\cos\theta|A||B|)^{\frac{P}{2}} + (|A|^2 + |B|^2 - 2\cos\theta|A||B|)^{\frac{P}{2}} \Big/ 2 \\ &\quad \left( t \rightarrow t^{\frac{P}{2}} \text{ convex} \right) \end{aligned}$$

get maximum when  $\cos\theta = 1 \Rightarrow \theta = 0$

Thm: (Projection onto closed convex sets in  $L^P$ )

Let  $(X, \mu)$  measure space,  $1 < P < \infty$ ,

$C \subseteq L^P$  convex closed.

(Important special case =  $C$  is closed linear subspace)

Let  $f \in L^P \setminus C$ , let  $d := \inf_{g \in C} \|f-g\|_P > 0$

Thm:  $\exists! g_0 \in C$ , s.t.  $\|f-g_0\|_P = d$ . "projection" of  $f$ .

Pf:  $2 \leq P < \infty$

Step 1: Construct a minimizing sequence  $\{g_k\}$  in  $C$ ,

i.e.  $\lim_{k \rightarrow \infty} \|f-g_k\|_P = d$

Step 2: Get a converged subsequence  $g_k \rightarrow g_0$ .

Step 3: Show  $\|f-g_0\|_P = d$

In our case, also to show uniqueness.

Given  $f \in L^P$ , by translating  $C$  to  $\{g-f \mid g \in C\}$ ,

we can take  $f=0$ .

Consider  $\{g_k\}$  minimizing sequence.  $\lim_{k \rightarrow \infty} \|g_k\|_P = d$

By H inf,  $\forall k, l \in \mathbb{N}$ ,

$$\begin{aligned} \|g_k + g_l\|_P^P + \|g_k - g_l\|_P^P &\leq (\|g_k\|_P + \|g_l\|_P)^P + (\|g_k\|_P - \|g_l\|_P)^P \\ \text{As } k, l \rightarrow \infty \xrightarrow{\| \cdot \|_P} &\xrightarrow{\| \cdot \|_P} \underbrace{\|g+g_0\|_P^P}_{\geq d^P} \xrightarrow{\downarrow} d^P \quad \downarrow \end{aligned}$$

So  $\downarrow$  this term sufficiently small  $\Rightarrow \{g_k\}$  cauchy.

$\Rightarrow \{g_k\}$  converges,  $\|f\|_P = \lim_{k \rightarrow \infty} \|g_k\|_P = d$ .

What can we say about the projection of  $f$  onto  $C$ ?

We know  $\|f-g\|_P^P \geq \|f-g_0\|_P^P, \forall g \in C$ .

In particular,  $\|f-(1-t)g_0+tg\|_P^P \geq \|f-g_0\|_P^P$  for  $0 \leq t \leq 1$

$$0 \leq \frac{d}{dt} \|f-(1-t)g_0+tg\|_P^P \Big|_{t=0} = \frac{d}{dt} \int |f-g_0+t(g-g_0)|^P dt$$

Differentiate the integrand

$$\begin{aligned} \frac{d}{dt} \left( |f-g_0+t(g-g_0)|^P \right)^{\frac{P}{2}} \Big|_{t=0} &= \frac{P}{2} \|f-g_0\|^{P-2} \frac{d}{dt} (f-g_0+t(g-g_0)) \overline{(f-g_0+t(g-g_0))} \Big|_{t=0} \\ &= \frac{P}{2} \|f-g_0\|^{P-2} \operatorname{Re}(g-g_0)(\overline{f-g_0}) \\ &= P \operatorname{Re}(\|f-g_0\|^{P-2} \overline{f-g_0})(g-g_0) \end{aligned}$$

New condition:

$\operatorname{Re} \int (|f-g_0|^{P-2} \overline{f-g_0})(g-g_0) d\mu > 0$ .  $\forall g \in C$ . Necessary condition for  $g_0$

Let  $C = V^\perp$  a closed vector subspace of  $L^P$ ,  $f \notin V$ .

$\int_V \operatorname{Re} \int (|f-g_0|^{P-2} \overline{f-g_0})(g-g_0) d\mu > 0, \forall f$

Let  $g \rightarrow g+g_0$   $\operatorname{Re} \int (|f-g_0|^{P-2} \overline{f-g_0})g d\mu > 0$

$g \rightarrow -g$   $\operatorname{Re} \int (|f-g_0|^{P-2} \overline{f-g_0})g d\mu = 0$

$g \rightarrow ig$   $\int (|f-g_0|^{P-2} \overline{f-g_0})g d\mu = 0$

Riesz Representation Thm:

$(X, \mu)$ , let  $1 \leq p < \infty$ . If  $P=1$ , assume  $\mu$  is  $\sigma$ -finite.

Then  $(L^P)^* = L^{P'}$

The canonical embedding  $L^P \hookrightarrow (L^P)^*$ ,  $g \mapsto T_g$

where  $T_g(f) = \int f g d\mu$ .

Pf: Let  $T \in (L^P)^*$ , i.e.,  $T: L^P \rightarrow \mathbb{C}$ , linear, cont.

Consider  $N = \{h \in L^P \mid T(h) = 0\}$ . ( $N$  is closed since  $T$  cont.)

If  $N = L^P$ , then  $T \geq 0$ ,

Otherwise, pick  $f_0 \in N$ . Take  $h_0 \in N$ , s.t.

$d(f_0, N) = \|f_0\|_{N^{\perp}}$ . By rescaling and translation we may

WLOG assume  $h_0 = 0$ ,  $\|f_0\|_N = 1$ . I.e. the projection of  $f_0$  on  $N$  is 0.

We proved that  $\int_{\text{constant}}^{f_0 \perp} \overline{\text{sign } f_0} h \, dh = 0, \forall h \in N$ .

Set  $g = C \int_{f_0 \perp}^{f_0 \perp} \overline{\text{sign } f_0}$

Then  $\int g h = 0, \forall h \in N$ .  $\int g f_0 = C \int_{f_0 \perp}^{f_0 \perp} \overline{\text{sign } f_0} f_0 \mid_{f_0 \perp} \text{sign } f = C \int f_0 \perp = C$ .

Set  $C = T f_0$ .

$\in N(T)$

$\in \text{span } f_0$

So  $\forall f \in L^P$ ,  $f = f - \frac{T f}{T f_0} f_0 + \frac{T f}{T f_0} f_0$

$$\begin{aligned} \int f g &= \int g \left( f - \frac{T f}{T f_0} f_0 \right) + \int g \left( \frac{T f}{T f_0} f_0 \right) \\ &= T f. \end{aligned}$$

So  $\exists q$ , s.t.  $T = T_q$ .  $\checkmark$

For  $P=1$ , firstly consider the case where  $X$  finite measure.

WLOG  $\mu(X)=1$ .

$$(p < q \Rightarrow L^q \subseteq L^P \Rightarrow (L^P)^* \subseteq (L^q)^*)$$

We already know that on  $L^P$  ( $1 < P < \infty$ ), the linear map  $T$  is represented by a unique elt  $v_p \in L^q$  w/  $\|v_p\|_{L^P} = \|T\|_{(L^P)^*} \leq \|T\|_{(L^q)^*}$ . Since  $L^q \subset L^P$  and  $L^q \supset L^P$  whenever  $p < q$ , we must have  $\forall p, q > 1, v_p = v_q \equiv v \in \bigcap_{1 < p < \infty} L^P$ .

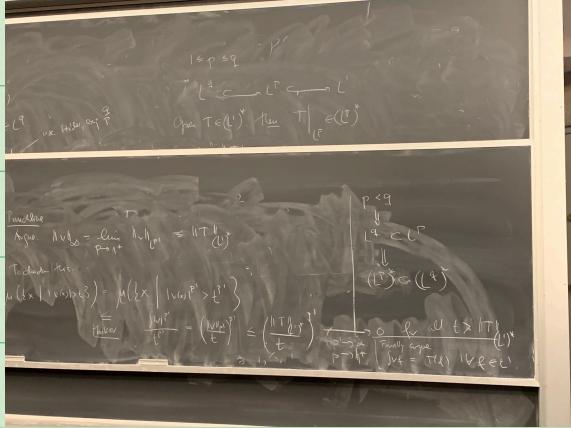
Punchline:

$$\text{Argue: } \|v\|_{L^P} = \lim_{p \rightarrow 1^+} \|v\|_{L^P} \leq \|T\|_{(L^P)^*}$$

To check this,

$$\mu(\{x\})$$

$$\begin{aligned} \text{That's } P < q &\quad \text{Punchline} \\ (\mathbb{I}_{\{x\}}) \subset L^P &\quad \text{Arg: } \|v\|_{L^P} = \lim_{p \rightarrow 1^+} \|v\|_{L^P} \leq \|T\|_{(L^P)^*} \\ \text{so } \text{for } f \in L^q &\quad \text{To check this:} \\ \text{we have } \|f\|_{L^P}^p &\leq \|f\|_{L^q}^q \quad \mu(\{x\} \cap \{v_n > t\}) \\ \text{by an element} &\quad \text{Punchline} \\ &\quad \text{Arg: } \|v\|_{L^P} = \lim_{p \rightarrow 1^+} \|v\|_{L^P} \leq \|T\|_{(L^P)^*} \\ &\quad \text{To check this:} \\ &\quad \mu(\{x \mid |v_n(x)|^p > t^p\}) \\ &\leq \frac{\|v\|_{L^P}^p}{t^p} = \left(\frac{\|v\|_{L^P}}{t}\right)^p \leq \left(\frac{\|T\|_{(L^P)^*}}{t}\right)^p \xrightarrow[p \rightarrow 1^+]{} 0 \quad \text{if all } t > 0 \\ &\quad \text{Finally argue: } \|v\|_{L^P} = \|T\|_{(L^P)^*} \end{aligned}$$

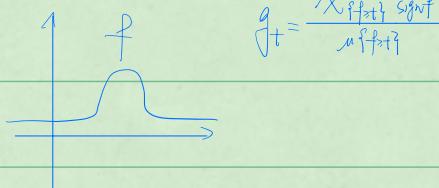


$$\bullet \quad \|f\|_p = \sup_{g \in L^p} |\int f g d\mu| = \sup_{g \in L^p} \operatorname{Re} \int f g d\mu \quad (1 < p < \infty)$$

$\underbrace{g = \frac{\operatorname{sign} f}{\|f\|_p^{1/p}}}_{\text{actually max.}}$

$$\bullet \quad p=1 \text{ case: } \|f\|_1 = \int |f| d\mu = \int |f| \operatorname{sign} f d\mu.$$

$$p=\infty \Rightarrow \|f\|_\infty = \sup_{\|g\|=1} |\int f g d\mu|$$



Consider  $1 \leq p_0 < p_1 < \infty$ ,  $p \in (p_0, p_1)$

Lemma:  $L^{p_0} \cap L^{p_1} \subseteq L^p \subseteq L^{p_0} + L^{p_1}$  (Continuous embeddings)

$$\text{Def: } L^{p_0} + L^{p_1} = \{f \mid f = f_0 + f_1, f_0 \in L^{p_0}, f_1 \in L^{p_1}\}.$$

$$\|f\|_{L^{p_0} + L^{p_1}} = \inf \{ \|f_0\|_{L^{p_0}} + \|f_1\|_{L^{p_1}} \}$$

$$\text{Rif: } L^{p_0} \cap L^{p_1} \subseteq L^p$$

$$p_0 < p < p_1 \Leftrightarrow \frac{1}{p_1} < \frac{1}{p} < \frac{1}{p_0}$$

$$\text{Find } t \in (0,1) \text{ s.t. } \frac{1}{p} = \frac{(1-t)}{p_0} + \frac{t}{p_1} \Rightarrow t = (1-t) \frac{p}{p_0} + t \frac{p}{p_1}$$

$$\text{Let } f \in L^{p_0} \cap L^{p_1}.$$

$$\begin{aligned} \|f\|_p^p &= \int |f|^p d\mu = \int |f_1|^{(1-t)p} |f_0|^{tp} d\mu \\ &\stackrel{\text{Holder}}{\leq} \left( \int |f_1|^{p_1} d\mu \right)^{\frac{(1-t)p}{p_1}} \left( \int |f_0|^{p_0} d\mu \right)^{\frac{tp}{p_0}} \\ &= \|f_1\|_{p_1}^{(1-t)p} \|f_0\|_{p_0}^{tp} \end{aligned}$$

$$\Rightarrow \|f\|_p \leq \|f_1\|_{p_1}^{(1-t)} \|f_0\|_{p_0}^t$$

NTs  $L^p \subseteq L^{p_0} + L^{p_1}$ . Given  $f \in L^p$ , need to decompose  $f = f_0 + f_1$ .

Take  $f_0 = (|f|-1)_+ \operatorname{sign} f$      $f_1 = \min(|f|, 1) \operatorname{sign} f$

$$\text{Markov's Ineq: } \mu \{ x \mid f(x) > t \} \leq \frac{\int |f|^p}{t^p} = \left( \frac{\|f\|_p}{t} \right)^p$$

$$\begin{aligned}
 \text{We estimate } \|f\|_{P_0}^{P_0} &= \int_0^\infty \mu \{x \mid |f(x)| > t\}^{\frac{P_0}{P_0-1}} dt \\
 &= \int_0^\infty \mu \{x \mid |f(x)| > t+1\}^{\frac{P_0}{P_0-1}} dt \\
 &\leq \int_0^\infty \|f\|_P^P \frac{t^{\frac{P_0-1}{P}}} {(t+1)^P} dt \\
 &\leq \int_0^\infty \|f\|_P^P t^{-1-\frac{P}{P_0}} dt < \infty
 \end{aligned}$$

$P_0, P_1$ . Determine  $P_t$  by  $\frac{1}{P_t} = \frac{1}{P_0} + \frac{t}{P_1}$

$$\text{Know: } \|f\|_{P_t} \leq \|f\|_{P_0}^{\frac{1}{P_0}} \|f\|_{P_1}^{\frac{t}{P_1}}$$

$$T: L^{P_0}_u + L^{P_1}_v \rightarrow L^{\frac{P_0}{P_0-1}} + L^{\frac{P_1}{P_1-1}}, \text{ s.t. } \|Tf\|_{L^{\frac{P_0}{P_0-1}}} \leq \|T\|_{P_0 \rightarrow \frac{P_0}{P_0-1}} \|f\|_{P_0}$$

$$\|Tf\|_{L^{\frac{P_1}{P_1-1}}} \leq \|T\|_{P_1 \rightarrow \frac{P_1}{P_1-1}} \|f\|_{P_1}$$

Riesz-Thorin Theorem:  $(X, \mathcal{M}), (\mathbb{Y}, \mathcal{L})$  semi-finite

Let  $P_0, P_1, q_0, q_1 \in [1, \infty]$  Fix  $t \in (0, 1)$  let  $P(t), q(t)$  be

$$\frac{1}{P(t)} = \frac{1-t}{P_0} + \frac{t}{P_1}, \quad \frac{1}{q(t)} = \frac{1-t}{q_0} + \frac{t}{q_1}$$

$$\text{Then } \|T\|_{L^{P(t)} \rightarrow L^{q(t)}} \leq \|T\|_{L^{P_0} \rightarrow L^{q_0}}^{1-t} \|T\|_{L^{P_1} \rightarrow L^{q_1}}^t$$

$$\|T\|_{P_0 \rightarrow q_0} = \sup \frac{\|Tf\|_{q_0}}{\|f\|_{P_0}} \quad \|Tf\|_{q(t)} \stackrel{\text{goal}}{\leq} \|Tf\|_{q_0}^{1-t} \|Tf\|_{q_1}^t$$

$$\|f\|_{P(t)} \stackrel{\text{def.}}{\leq} \|f\|_{P_0}^{1-t} \|f\|_{P_1}^t$$

Key observation:  $t \rightarrow \left( \int_0^1 \frac{1}{P(s)} ds \right)^{\frac{1}{P(t)}}$ ,  $t \rightarrow at = e^{t \log a}$

$$P(t) = \left( \frac{1-t}{P_0} + \frac{t}{P_1} \right)^{-1} \rightarrow \text{analytic holomorphic over } \mathbb{C}$$

3-line lemma: Let  $S = \{z \in \mathbb{C} \mid 0 < \operatorname{Re}(z) < 1\}$

$\Psi: S \rightarrow \mathbb{C}$  holomorphic. cont on  $\bar{S}$

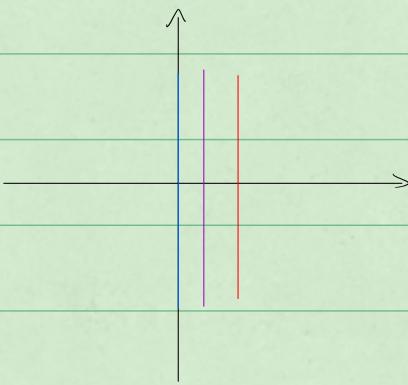
Assume that  $|\Psi(z)| \leq M_0$ , if  $\operatorname{Re}(z) = 0$

$|\Psi(z)| \leq M_1$ , if  $\operatorname{Re}(z) = 1$

Then  $\sup_{\operatorname{Re}(z)=t} |\Psi(z)| \leq M_0^t M_1^{1-t}$

PF: WLOG  $M_0 = M_1 = 1$ , otherwise consider  $\Psi(z) = \frac{\Psi(z)}{M_0^z M_1^z}$

Consider  $\tilde{\Psi}(z) = \Psi(z) \cdot \exp(\varepsilon(z^2 - 1))$



Note  $|\tilde{\Psi}(z)| \leq |\Psi(z)|$  holomorphic,  $|\tilde{\Psi}(z)| \rightarrow 0$  as  $|z| \rightarrow \infty$  on  $S$

$$\tilde{\Psi}(z) \xrightarrow{z \rightarrow \infty} \Psi(z) \text{ pointwise.}$$

By maximum principle,  $\sup_z |\tilde{\Psi}(z)| \leq$

Pf. Given  $1 < p_0, p_1, q_0, q_1 < \infty$ . T.

$T|_{L^{p_0}}: L^{p_0} \rightarrow L^{q_0}$ ,  $T|_{L^{p_1}}: L^{p_1} \rightarrow L^{q_1}$ , bounded linear.

$$WLS \quad \|T\|_{L^{p_0} \rightarrow L^{q_0}} \leq \|T\|_{L^{p_0} \rightarrow q_0}^{1-t} \|T\|_{L^{p_1} \rightarrow q_1}^t$$

$$\|T\|_{L^{p_0} \rightarrow L^{q_0}} = \sup_{\substack{f \in L^{p_0} \\ \|f\|_{p_0}=1}} \|Tf\|_{q_0}$$

$$\stackrel{\text{defn}}{=} \sup_{\substack{f \in L^{p_0} \\ \|f\|_{p_0}=1}} \sup_{\substack{g \in L^{q_1} \\ \|g\|_{q_1}=1}} \left| \int_T f g d\mu \right|$$

$$WLOG \quad f = \sum_i^n a_i \chi_{E_i}, \quad g = \sum_j^m b_j \chi_{F_j}$$

Interpolation: For  $0 \leq \operatorname{Re} z \leq 1$ ,  $\frac{1}{p(z)} = \frac{1-z}{p_0} + \frac{z}{p_1}$

$$f_z := |f|^{\frac{p(z)}{p_0}} \operatorname{sign} f, \quad g_z := |g|^{\frac{q(z)}{q_1}} \operatorname{sign} g \quad (g_1 = g, f_1 = f)$$

Clearly  $z \mapsto \int T(f_z) g_z$  is holomorphic on  $S = \{0 \leq \operatorname{Re} z \leq 1\}$ ,

continuous on  $\overline{S}$ .

$$\text{Write } \int T(f) g d\mu = \int T\left(\sum_i^n a_i \chi_{E_i}\right) \left(\sum_j^m b_j \chi_{F_j}\right) d\mu$$

$$= \sum_i \sum_j a_i b_j \int T(\chi_{E_i}) (\chi_{F_j}) d\mu$$

$$\text{Note: } f_1 = f, \quad \|f_1\|_{p_0} = \int |f|^{\frac{p_1}{p_0}} d\mu = \|f\|_{\frac{p_1}{p_0}} = 1$$

$$\text{Similarly, } \|f_z\|_{p_0} = \|g_z\|_{q_1} = \|g\|_{q_1} = 1$$

$$\|f_z\|_{p_0} = \sqrt{\int |f|^{\frac{p_1}{p_0}} d\mu} = 1$$

$$\text{For } z=i, \quad \int T(f_z) g_z d\mu \leq \|T\|_{L^{p_0} \rightarrow L^{q_0}} \|f_z\|_{p_0} \|g_z\|_{q_1}$$

$$\Psi(z) = \int T(f_z) g_z d\mu, \quad \Psi(i) = \int T(f_i) g_i d\mu \leq \|T\|_{L^{p_0} \rightarrow L^{q_0}} \|f_i\|_{p_0} \|g_i\|_{q_1}$$

$$3-\text{line lemma: } |\Psi(z)| \leq \|T\|_{L^{p_0} \rightarrow L^{q_0}}^{1-t} \|T\|_{L^{p_1} \rightarrow L^{q_1}}^t$$

Let  $f = \mu$ -measurable function, complex valued.

The level set of  $|f|$  at height  $\alpha$  is given by

$$E_f(\alpha) = \{x \in X \mid |f(x)| > \alpha\},$$

Distribution function of  $f$ :

$$\lambda_f(\alpha) = \mu(E_f(\alpha))$$

$$\text{Recall: } \|f\|_P^P = \int_0^\infty \lambda_f(\alpha) P \alpha^{P-1} d\alpha$$

$$\text{Markov inequality: } \mu(E_\alpha) \leq \left(\frac{\|f\|_P}{\alpha}\right)^P$$

$$\text{Def: } E_f^P := \inf \{C \in \mathbb{R} \mid \lambda_f(\alpha) \leq \left(\frac{C}{\alpha}\right)^P\} = \sup \alpha \mid (\lambda_f(\alpha))^{\frac{1}{P}}$$

weak- $L^P = \{f \text{ measurable, } \|f\|_P < \infty\} / \text{u.a.e.}$

By Markov inequality, weak- $L^P \subseteq L^P$

Note:  $[cf]_P = |c| E_f^P$ , but  $\|\cdot\|$  is not a norm.

$$\lambda_{f+g}(\alpha) = \mu\{|f(x)+g(x)| > \alpha\} \leq \mu\{|f(x)| > \alpha \text{ or } |g(x)| > \alpha\} \leq \lambda_f(\alpha) + \lambda_g(\alpha)$$

$$E_{f+g}^P \leq E_f^P + E_g^P$$

Def:  $f_n \rightarrow f$  in weak- $L^P \Leftrightarrow \lim E_{f_n}^P - E_f^P = 0$

Lemma: If  $1 \leq p_0 < p < p_1 < \infty$ , then  $(\text{weak-}L^{p_0}) \cap (\text{weak-}L^{p_1}) \subseteq L^P$

Example:  $f \in L' \Rightarrow$  Hardy-Littlewood  $\Rightarrow$   $\max f_n$  lies in  $\text{weak-}L'$

$Pf$ : Assume  $C_0, C_1$  constant, s.t.  $\lambda_f(\alpha) \leq (\frac{C_0}{\alpha})^{p_0}$ ,  $\lambda_f(\alpha) \leq (\frac{C_1}{\alpha})^{p_1}$ ,  $\forall \alpha > 0$

Write  $\frac{1}{P} = \frac{1-t}{p_0} + \frac{t}{p_1}$ , take  $a > 0$

$$\text{We estimate } \|f\|_P^P = \int_0^\infty \lambda_f(\alpha) P \alpha^{P-1} d\alpha$$

$$= \int_0^a \lambda_f(\alpha) P \alpha^{P-1} d\alpha + \int_a^\infty \lambda_f(\alpha) P \alpha^{P-1} d\alpha$$

$$\leq \int_0^a \left(\frac{C_0}{\alpha}\right)^{p_0} P \alpha^{P-1} d\alpha + \int_a^\infty \left(\frac{C_1}{\alpha}\right)^{p_1} P \alpha^{P-1} d\alpha$$

$$= C_0^P \frac{P}{P-p_0} a^{P-p_0} + C_1^P \frac{P}{P-p_1} a^{P-p_1}$$

$$\min_{\alpha} \frac{C_0^P C_1^P}{(1-t)^{P-p_0} + t^{P-p_1}}$$

Consider a map  $T: L^p \rightarrow \text{weak-}L^q$

We say  $T$  is **sublinear** if

$$(1) |T(f+g)| \leq |Tf| + |Tg| \quad (\text{pointwise})$$

$$(2) |T(cf)| = |c| |Tf| \quad (\text{pointwise})$$

We say  $T$  is **strong-type**  $(p, q)$  if  $\|Tf\|_q \leq \text{constant} \|f\|_p$   
**weak-type**  $(p, q)$  if  $|Tf|_q \leq \text{constant} \|f\|_p$

Thm: (M real-interpretation)

Let  $1 \leq p_0, p_1, q_0, q_1 < \infty$ ,  $f_0 \neq f_1$ ,  $p_0 < q_0$ ,  $p_1 < q_1$ ,  $0 < t < 1$

Let  $L^{p_0} + L^{p_1} \rightarrow (\text{weak } L^{q_0}) + (\text{weak } L^{q_1})$  be sublinear.

Assume  $T$  is weak-type  $(p_0, q_0)$  and weak-type  $(p_1, q_1)$

$$\text{Define } \frac{1}{p} = \frac{1-t}{p_0} + \frac{t}{p_1}, \quad \frac{1}{q} = \frac{1-t}{q_0} + \frac{t}{q_1}$$

Then  $T$  is strong-type  $(p, q)$

Pf: Assume  $p_0 < p_1$ , ( $p_0 = p_1$  case equivalent to the lemma)

For  $a > 0$ , write  $f = f^a + f_a$ , where  $f^a = (|f| - a) \text{sign} f$ ,  $f_a = \min(|f|, a) \text{sign} f$

$$\lambda_{f^a}(x) = \lambda_{f^a}(a+x) \leq \lambda_f(x), \quad \lambda_{f_a}(x) = \begin{cases} \lambda_f(x) & x \leq a \\ 0 & x > a. \end{cases}$$

$$\lambda_T f(x) \leq \lambda_{Tf^a}(x) + \lambda_{Tf_a}(x)$$

$$\text{We write } \|Tf\|_q^q = \int_0^\infty \lambda_T f(x)^q x^{q-1} dx$$

$$\leq 2^q \int_0^\infty (\lambda_{Tf^a}(x) + \lambda_{Tf_a}(x))^q x^{q-1} dx$$

$$\leq 2^q \int_0^\infty \left( \left( \frac{|Tf^a|}{x} \right)^{\frac{q_0}{p_0}} + \left( \frac{|Tf_a|}{x} \right)^{\frac{q_1}{p_1}} \right)^q x^{q-1} dx$$

$$\leq 2^q \int_0^\infty \left( \left( \frac{C_0 \|f\|_p}{x} \right)^{\frac{q_0}{p_0}} + \left( \frac{C_1 \|f\|_p}{x} \right)^{\frac{q_1}{p_1}} \right)^q x^{q-1} dx \quad \text{Since } \|Tf\|_q \leq C \|f\|_p$$

Markov

$$\leq 2^q C_0^q \int_0^\infty \int_\alpha^\infty \lambda_{f^a}(\beta) \beta^{\frac{p_0-1}{p_0}} \frac{x^{\frac{q_0}{p_0}}}{\beta^{\frac{p_0}{p_0}}} x^{q-1} d\beta dx$$

$$+ 2^q C_1^q \int_0^\infty \int_0^{\alpha} \lambda_{f_a}(\beta) \beta^{\frac{p_1-1}{p_1}} \frac{x^{\frac{q_1}{p_1}}}{\beta^{\frac{p_1}{p_1}}} x^{q-1} d\beta dx$$

Key idea: we can take  $\alpha$  to depend on  $x$ .

$$a = x^\sigma \quad (1) \text{ integral: } 0 \leq \beta \leq x^\sigma \\ (2) \dots \quad x \leq \beta < \infty \quad \sigma, \text{TBD.}$$

First integral = Apply Minkowski's inequality with  $r_0 = \frac{P_0}{q_0}$  and integrand

教授挂黑板了...

Important example of real interpolation.

Consider  $L^1(\mathbb{R}, m)$

Lebesgue differential thm

Assume  $f: \Omega \rightarrow \mathbb{R}$  integrable. Then

$$\int_{B_r(x)} f(y) dy = \frac{1}{m(B_r(x))} \int_{B_r(x)} f$$

$$\lim_{r \rightarrow 0} \int_{B_r(x)} f(y) dy = f(x) \text{ for a.e. } x.$$

Proof using Hardy-Littlewood maximal fn.

$$Hf(x) = \sup_{r>0} \int_{B_r(x)} f(y) dy$$

Today's Thm: If  $f \in L^p(\mathbb{R}^n)$ , then  $Hf \in L^p(\mathbb{R}^n)$  for  $1 < p < \infty$

And  $\|Hf\|_p \leq C_n \|f\|_p$ .

Proof: Interpolation between  $p=1$ ,  $p=\infty$ .

$$p=\infty = \|Hf\|_\infty = \sup_{x \in \mathbb{R}^n} \sup_r \int_{B_r(x)} |f(y)| dy \leq \|f\|_\infty$$

$p=1 = \text{Hardy-Littlewood maximal thm: } \exists \text{ constant } C_n \leq 3^n,$

$$\text{s.t. } \forall \alpha > 0, \quad m\{|x| > Hf(x) > \alpha\} \leq \frac{C_n \|f\|_1}{\alpha}$$

So  $Hf$  is weak-type  $(1,1)$ , strong-type  $(\infty, \infty)$  *Therefore weak*.

Is  $H$  sublinear?

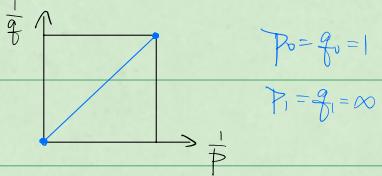
$$H_cf(x) = \sup_r \int_{B_r(x)} |cf(y)| dy = c H_f(x)$$

$$\begin{aligned} H_{f+g}(x) &= \sup_r \int_{B_r(x)} |H_f(y) + g(y)| dy \\ &\leq \sup_r \int_{B_r(x)} |H_f(y)| + |g(y)| dy \end{aligned}$$

$$\leq \sup_r \int_{B_r(x)} |H_f(y)| dy + \sup_r \int_{B_r(x)} |g(y)| dy$$

$$= H_f(x) + H_g(x)$$

So  $H$  is sublinear, weak-type  $(1,1), (\infty, \infty)$



## Hilbert Space

Def: A Hilbert Space is a vector space with a norm given by inner product. The vector space is assumed to be complete with that norm.

$$\mathbb{L}^2(\Omega), \langle f, g \rangle = \int f \bar{g} d\mu$$

Let  $V$  be a vector space

Def: An inner product on  $V$  is a map:  $\langle \cdot, \cdot \rangle$  from  $V \times V$  to  $\mathbb{C}$  such that

①  $\langle x, x \rangle \geq 0$  with  $= \Leftrightarrow x = 0$ , positivity

②  $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle$ , linearity

③  $\langle x, y \rangle = \overline{\langle y, x \rangle}$ , Hermitian.

Norm is  $\|x\| = \sqrt{\langle x, x \rangle}$

Bessel inequality:  $|\langle x, y \rangle| \leq \sqrt{\|x\|^2 + \|y\|^2}$ ,  $|\langle x, y \rangle| \leq \|x\| \|y\|$ .

Triangular inequality  $\|x+y\| \leq \|x\| + \|y\|$

Parallelogram Identity:

$$\|x+y\|^2 + \|x-y\|^2 = 2(\|x\|^2 + \|y\|^2)$$

Consequences: If  $H$  a Hilbert Space,  $C \subseteq H$  closed, convex,

$x \notin C$ , then  $\exists! y_0 \in C$  s.t.  $\|x - y_0\| = \inf_{y \in C} \|x - y\|$   $y_0 = P_C(x)$

Necessary condition on  $y_0 = \operatorname{Re} \langle x - y_0, y - y_0 \rangle \geq 0, \forall y \in C$ .

If  $C$  subspace,  $\langle x - y_0, y - y_0 \rangle = 0, \forall y \in C$ .

Def:  $x, y$  are orthogonal if  $\langle x, y \rangle = 0$ .

The angle between  $x, y$  are  $\cos \alpha = \frac{\langle x, y \rangle}{\|x\| \|y\|}$

An orthonormal sequence in  $H$  is a sequence  $\{u_k\}_{k \geq 1}$  s.t.  $\langle u_k, u_j \rangle = \delta_{ij}$ .

Remarks: If  $x_1, \dots, x_n$  mutually orthogonal, then

$$\|x_1 + \dots + x_n\|^2 = \sum \|x_i\|^2.$$

Let  $V \subseteq H$  closed subspace.  $x \notin V, \exists y_0 \in V, \|x - y_0\| = \inf_V \|x - y\|$

$$\Rightarrow (x - y_0) \perp y, \forall y \in V$$

Is the converse true? Yes.

Assume  $\langle x - y_0, y \rangle = 0, \forall y \in V$ .

$$\|x - y\|^2 = \| (x - y_0) - (y - y_0) \|^2 = \|x - y_0\|^2 + \|y - y_0\|^2 \geq \|x - y_0\|^2.$$

Def:  $V \subseteq H$  linear subspace, its orthogonal complement is

$$V^\perp = \{x \in H \mid \langle x, y \rangle = 0, \forall y \in V\}.$$

Fact:  $V$  closed  $(V^\perp)^\perp = \overline{V}$ .

Excision: Infinite sum.

Let  $N$  be a set,  $\{a_n\}_{n \in N}$  a family of non-negative numbers.

$$\text{Def: } \sum_{n \in N} a_n = \sup_{F \subseteq N \text{ finite}} \sum_{n \in F} a_n$$

$$\text{Suppose } \sum_{n \in N} a_n < \infty$$

Claim: Then  $\{n \in N \mid a_n > 0\}$  is countable.

## Bessel's Inequality

Assume  $(v_n)_{n \in N}$  orthonormal family, then  $\forall x \in H$ , we have

$$\sum_{n \in N} |\langle x, v_n \rangle|^2 \leq \|x\|^2$$

Consider first a finite sequence  $v_1, \dots, v_n$  orthonormal.

Given  $x$ , consider  $x - \sum_{j=1}^n \langle x, v_j \rangle v_j$

$$V = \text{span}\{v_1, \dots, v_n\}. \quad \text{Claim: } \sum_{j=1}^n \langle x, v_j \rangle v_j = P_V x, \text{ since } \langle x - \sum_{j=1}^n \langle x, v_j \rangle v_j, v_k \rangle = 0$$

$$\text{We conclude } \|x\|^2 = \|x - \sum_{j=1}^n \langle x, v_j \rangle v_j + \sum_{j=1}^n \langle x, v_j \rangle v_j\|^2 \geq \sum_{j=1}^n |\langle x, v_j \rangle|^2 \quad (\text{Equality } \Leftrightarrow x \in V)$$

$$\Rightarrow \sum_{n \in N} |\langle x, v_n \rangle|^2 \leq \sum_{\text{finite}} |\langle x, v_n \rangle|^2 \leq \|x\|^2$$

Thm: Let  $H$  be an infinite-dim separable Hilbert space, and let

$(v_n)_{n \in N}$  orthonormal sequence. TFAE

① Density:  $\text{Span}\{v_{n \in N}\}$  dense in  $H$

② Completeness:  $\forall x \in H$ , if  $\langle x, v_n \rangle = 0 \quad \forall n$ , then  $x = 0$

③ Equality in Bessel:  $\forall x \in H, \|x\|^2 = \sum_{n=1}^{\infty} |\langle x, v_n \rangle|^2$

④ Representation:  $\forall x \in H, \exists (a_n)_{n \in N} \subseteq \mathbb{C}$ , s.t.

$$x = \lim_{N \rightarrow \infty} \sum_{n=1}^N a_n v_n = \sum_{n=1}^{\infty} a_n v_n.$$

Rf: ①  $\Rightarrow$  ②  $\Rightarrow$  ③  $\Rightarrow$  ④  $\Rightarrow$  ①

①  $\Rightarrow$  ② = Assume span is dense,

Let  $x \in H$  be given. If  $\langle x, v_n \rangle = 0, \forall n \in N$ ,

then  $\langle x, y \rangle = 0, \forall y \in \text{span} \Rightarrow \forall y \in \overline{\text{span}}$  since  $\langle \cdot, \cdot \rangle$  cont.

$\Rightarrow \forall y \in \overline{\text{Span}} = H$  (dense)  $\Rightarrow \langle x, y \rangle = 0 \Rightarrow x = 0$

(2)  $\Rightarrow$  (3) Consider  $S_N = \lim_{N \rightarrow \infty} \sum_{n=1}^N \langle x, v_n \rangle v_n$ . Claim:  $S_N$  converges in  $H$ .

$$\|S_N - S_{N+k}\|^2 = \left\| \sum_{n=N+1}^{N+k} \langle x, v_n \rangle v_n \right\|^2 = \frac{1}{N+k} \left( \sum_{n=1}^{N+k} |\langle x, v_n \rangle|^2 \right) \xrightarrow{N \rightarrow \infty} 0 \quad \text{Since } \sum_{n=1}^{\infty} |\langle x, v_n \rangle|^2 < \infty$$

So  $S := \sum_{n=1}^{\infty} \langle x, v_n \rangle v_n$  converges in  $H$

$$\text{Check } \langle x - S, v_k \rangle = \langle x - \sum_{n=1}^{\infty} \langle x, v_n \rangle v_n, v_k \rangle$$

$$= \langle x, v_k \rangle - \sum_{n=1}^{\infty} \langle x, v_n \rangle \langle v_n, v_k \rangle$$

$$= 0 \Rightarrow x = S \text{ by (2)}$$

$$\Rightarrow x = \sum_{n=1}^{\infty} \langle x, v_n \rangle v_n \Rightarrow \|x\|^2 = \sum \langle x, v_n \rangle$$

(3)  $\Rightarrow$  (1) = Set  $a_n = \langle x, v_n \rangle$ .  $S := \sum_{n=1}^{\infty} \langle x, v_n \rangle v_n$  if absolute convergent

in  $H$  by Bessel)

$$\|x - S\|^2 = \lim_{N \rightarrow \infty} \|x - \sum_{n=1}^N \langle x, v_n \rangle v_n\|^2 = \lim_{N \rightarrow \infty} \left( \|x\|^2 - \left\| \sum_{n=1}^N \langle x, v_n \rangle v_n \right\|^2 \right) \stackrel{(3)}{\equiv} 0$$

Example:  $H = L^2(-\pi, \pi)$ , measure  $\frac{1}{2\pi} dx$

Standard Fourier basis:  $\{e_n = e^{inx}\}$

$$\langle e_n, e_m \rangle = \int_0^{2\pi} e^{inx} e^{-imx} = \int_0^{2\pi} e^{i(n-m)x} = \delta_{mn}$$

Fourier coefficient: given  $f \in L^2(-\pi, \pi)$ ,  $a_n = \hat{f}(n) = \int_{-\pi}^{\pi} f(x) e^{inx} dx$

is  $\{e_n\}$  an orthonormal basis? Yes, by showing density.

by using trig polynomial.  $\sum a_n e^{inx}$

The standard Fourier basis

Then  $\{e_n(x)\}$  is an orthonormal basis of  $H$ .

$\mathbb{R}^{\mathbb{Z}/N\mathbb{Z}}$   $\text{span}\{e_n\}$  is dense.

Consider for the moment,  $f: \mathbb{R} \rightarrow \mathbb{R}$ ,  $2\pi$ -periodic, continuous.

Clearly,  $f|_{(-\pi, \pi)} \in H$ , Define  $\hat{f}(k) = \langle f, e_k \rangle = \int_{-\pi}^{\pi} f(x) e^{-ikx} dx$

Consider  $F_f(x) = \sum_{k=-\infty}^{\infty} |r|^k \hat{f}(k) e^{ikx} \quad r \in \mathbb{C}, x \in \mathbb{R}$ .

$$= \sum_{k=-\infty}^{\infty} |r|^k \int_{-\pi}^{\pi} e^{ik(x-y)} dy$$

$$= \int_{-\pi}^{\pi} f(y) \left( \sum_{k=-\infty}^{\infty} r^{|k|} e^{ik(x-y)} \right) dy$$

Compute  $\Pr_r(x) = \sum_{k=-\infty}^{\infty} r^{|k|} e^{ikx}$

$$= \sum_{k \geq 0}^{\infty} (re^{ix})^k + \sum_{k < 0}^{\infty} (re^{-ix})^k - 1$$

$$= \frac{1}{1-re^{ix}} + \frac{1}{1-re^{-ix}} - 1$$

$$= \frac{1-re^{ix}+1-re^{-ix}}{(1-r\cos x)^2 + (r\sin x)^2} - 1$$

$$\text{So } \Pr_r(x) = \frac{2-2r\cos x - (1+r^2-2r\cos x)}{1+r^2-2r\cos x}$$

$$= \frac{1-r^2}{1+r^2-2r\cos x} - \frac{1}{2\pi}$$

$$\text{So } F_r(x) = \int_{-\pi}^{\pi} f(y) \Pr_r(x-y) dy \quad \text{Convolution!}$$

$$\text{WTS } \lim_{r \rightarrow 1^-} F_r(x) = f(x), \forall x.$$

$$\textcircled{1} \int_{-\pi}^{\pi} \Pr_r = 1 \quad \textcircled{2} \Pr_r \geq 0$$

$$\textcircled{3} \lim_{r \rightarrow 1^-} \Pr_r(x) = 0 \text{ uniformly on } (-\pi, \pi) \setminus (-\delta, \delta) \forall \delta.$$

Let  $\delta > 0$  TBD.

We estimate

$$|F_r(x) - f(x)| = \left| \int_{-\pi}^{\pi} \Pr_r(x-y) f(y) dy - f(x) \right|$$

$$= \left| \int_{-\pi}^{\pi} \Pr_r(x-y) (f(y) - f_w) dy \right|$$

$$\leq \int_{-\pi}^{\pi} \Pr_r(x-y) (|f(x)| + |f(y)|) \chi_{|x-y| \in (\delta, \pi)} dy + \int \chi_{|x-y| \notin (\delta, \pi)} |\Pr_r(x-y) (f(x) - f(y))| dy.$$

Given  $\varepsilon > 0$ , pick  $\delta$  s.t.  $|x-y| < \delta \Rightarrow |f(x) - f(y)| < \frac{\varepsilon}{3}$

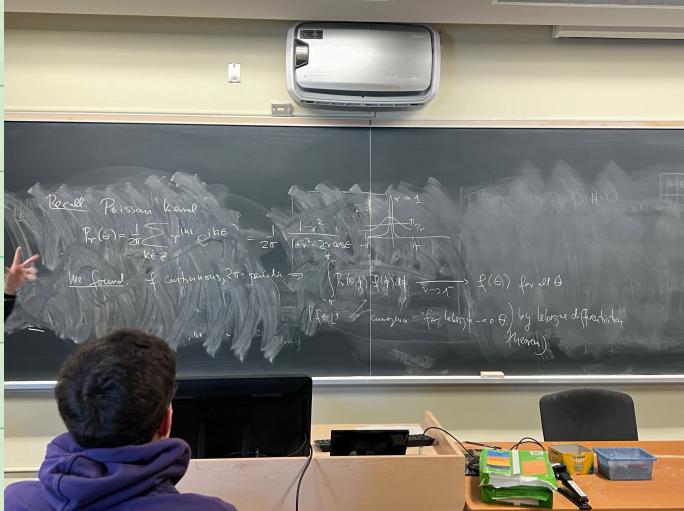
Pick  $r$  close to 1 s.t.  $|\Pr_r(x-y)| < \frac{\varepsilon}{3}$

$$\int_{-\pi}^{\pi} \Pr_r(x-y) (|f(x)| + |f(y)|) \chi_{|x-y| \in (\delta, \pi)} dy + \int \chi_{|x-y| \notin (\delta, \pi)} \underbrace{\Pr_r(x-y)}_{\leq 1} \underbrace{|\Pr_r(x-y) (f(x) - f(y))|}_{\leq \frac{\varepsilon}{3}} dy$$

$$\leq \frac{\varepsilon}{3} + \frac{\varepsilon}{3} \leq \frac{2}{3} \varepsilon.$$

Fix  $r$  s.t. above estimation holds.

$F_r(x) = \sum_{k \in \mathbb{Z}} r^{|k|} \hat{f}(k) e^{ikx}$ . Choose  $N$  s.t.  $|F_r(x) - \sum_{|k| \leq N} r^{|k|} \hat{f}(k) e^{ikx}| < \frac{\varepsilon}{3}$  ✓



Q. Consider  $f$  as a function  $D \rightarrow \mathbb{C}$ , Assume  $f$  is real.

Define:  $U(r\cos\theta, r\sin\theta) = \int_{-\pi}^{\pi} \Pr(\theta - \psi) f(\psi) d\psi$ . in unit disk.

$$r^k e^{ik\theta} = (re^{i\theta})^k = z^k \quad \text{holomorphic}$$

$$k < 0 : \quad r^{|k|} e^{-ik\theta} = (re^{-i\theta})^{|k|} = (\bar{z})^{|k|}$$

$$\text{Equivalently, } U(z) = \sum_{k \geq 0} \hat{f}(k) z^k + \sum_{k > 0} \hat{f}(-k) (\bar{z})^k + \hat{f}(0)$$

extending from  $\mathbb{S}^1$  to  $D$ .

Fact: A real-valued fn  $g: D \rightarrow \mathbb{R}^2$  is harmonic ( $\Delta g = 0$ )

$\Leftrightarrow g = \operatorname{Re} \Phi$ ,  $\Phi$  holomorphic.

$u$  solves the boundary-value problem  $\begin{cases} \Delta u = 0 \text{ on } D \\ u|_{\partial D} = \varphi \end{cases}$

We proved:  $\hat{f} \in L^2(-\pi, \pi) \Rightarrow \int_{-\pi}^{\pi} |f(x)|^2 dx = \sum_{k \in \mathbb{Z}} |\hat{f}(k)|^2$ . (i.e.  $\{\hat{f}(k)\} \in l^2$ )

injective ✓. Surjective? Yes.

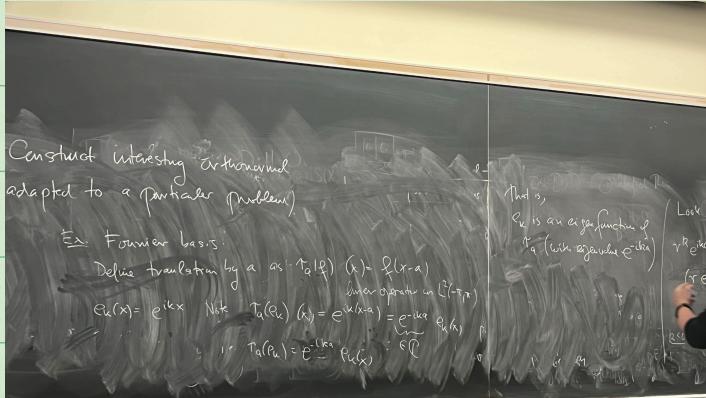
Q: Given a sequence  $(a_k)_{k \in \mathbb{Z}}$ , s.t.  $\sum a_k$  summable.

Define:  $f(x) := \sum_k a_k e^{ikx}$ , converges in  $L^2$ . pointwise convergent?

Consider partial sum  $S_N(x) = \sum_{k=-N}^N a_k e^{ikx}$

General theorem:  $\|f_n \rightarrow f\|_{L^2} \Rightarrow f_n \rightarrow f$  p.w. a.e.

Goal: Construct intersecting orthonormal basis



Finite dim case: If  $A \in \mathbb{C}^{n \times n}$  is Hermitian ( $\bar{A}^t = A$ ), then  $\mathbb{C}^n$  has a basis of orthonormal eigenvectors.

Ex: no eigenvector:

$$T: \ell^2 \rightarrow \ell^2, T(a_1, a_2, a_3, \dots) = (0, a_1, a_2, \dots)$$

Def: Let  $H_1, H_2$  Hilbert Spaces,

$T: H_1 \rightarrow H_2$  a bounded linear operator

The adjoint of  $T$  is the operator  $T^*: H_2 \rightarrow H_1$ , s.t.

$$\langle Tx, y \rangle_{H_2} = \langle x, T^*y \rangle_{H_1}.$$

We say  $T: H \rightarrow H$  self-adjoint if  $T^* = T$

Does  $*$  uniquely determine  $T^*$ ?

Given  $y$ , consider the map  $L_y: x \mapsto \langle Tx, y \rangle \in \mathbb{C}$

$$L_y(H)^* \quad \left| \frac{|Tx|}{\|x\|_{H_1}} \leq \frac{\|Tx\|_{H_2} \|y\|_{H_2}}{\|x\|_{H_1}} \leq \|T\|_{H_1 \rightarrow H_2} \|y\|_{H_2} \right)$$

$$L_y(x) = \langle x, z \rangle \text{ for some } z \in \mathbb{C}, T^*y = z.$$

$$\text{Unique? } \langle Tx, y \rangle = \langle x, Ay \rangle = \langle x, By \rangle$$

$$\Rightarrow \langle x, (A-B)y \rangle = 0, \forall x, y \Rightarrow A=B$$

Norm of  $T^*$ ?  $\|T^*y\|_{H_1} = \sup_{\|x\|_{H_1}=1} \operatorname{Re}(\langle x, T^*y \rangle)$

$$\|T^*\| = \sup \frac{\operatorname{Re} \langle x, T^*y \rangle}{\|x\|_{H_1} \|y\|_{H_2}} = \sup \frac{\operatorname{Re} \langle Tx, y \rangle}{\|x\|_{H_1} \|y\|_{H_2}} = \|T\|$$

Remark:  $(T^*)^* = T$

$$\langle T^*x, y \rangle = \langle x, T^*y \rangle = \overline{\langle T^*y, x \rangle} = \overline{\langle y, T^{**}x \rangle}$$

Excercise:  $H^* = H$ ?

Given  $y \in H$ , can define  $L_y \in H^*$  s.t.  $L_y(x) = \langle x, y \rangle$

Given  $y \in H$ , can define  $L_y \in H^*$  by  
 $L_y(x) = \langle x, y \rangle$ . bounded linear  
 $\|L_y\| = \|y\|$  by Schwarz  
 $L_{y_1+y_2}(x) = \langle x, y_1+y_2 \rangle = \langle x, y_1 \rangle + \langle x, y_2 \rangle = L_{y_1}(x) + L_{y_2}(x)$   
 $L_{\alpha y}(x) = \langle x, \alpha y \rangle = \bar{\alpha} \langle x, y \rangle = \bar{\alpha} L_y(x)$   
 $y \rightarrow L_y$  not linear but  
 $H \rightarrow H^*$  not linear

Spectral Thm:

Def: Let  $X, Y$  Banach spaces,  $L: X \rightarrow Y$  bounded, linear.

$L$  is compact if every bounded sequence  $(x_n)$  in  $X$

the sequence  $(Lx_n)$  has a convergent subsequence.

Thm: Let  $H$  be separable Hilbert Space,  $A: H \rightarrow H$  compact,

self-adjoint then  $\exists (u_n)$  orthonormal basis,  $(\lambda_n)$  in  $\mathbb{R}$ ,

s.t.  $Au_n = \lambda_n u_n$

Strategy: ① Construct one eigenvalue/eigenvector pair.

Case 1:  $\lambda = \|A\|$  an eigenvalue.

$$\lambda = \max_{\|x\|=1} \langle x, Ax \rangle$$

Case 2:  $\lambda = -\|A\|$ , replace  $A$  by  $-A$

② Induction: Assume we have  $\{u_1, \dots, u_n\}, \{\lambda_1, \dots, \lambda_n\}$

orthonormal eigenvectors.

Restrict A to  $\{u_1, \dots, u_n\}^\perp$

③ Error estimate  $\|P_{\{u_1, \dots, u_n\}^\perp} AP_{\{u_1, \dots, u_n\}^\perp}\| \leq \|A_{n+1}\|$

Re: Step 1:  $\|A\| = \sup_{\|x\|=1} \|Ax\| = \sup_{\|y\|=1} \sup_{\|x\|=1} |\langle Ax, y \rangle| \leq \frac{1}{2} \operatorname{Re}(\langle A(x+y), x+y \rangle - \langle A(x-y), x-y \rangle + i \langle A(x+iy), x+iy \rangle - i \langle A(x-iy), x-iy \rangle)$

Polarization

$$\leq \frac{1}{4} \left| \|A\| \left( (\|x\| + \|y\|)^2 + (\|x\| - \|y\|)^2 \right) \right| = \frac{1}{2} \|A\| (\|x\|^2 + \|y\|^2)$$

Purely imaginary

Claim:  $\|A\| = \sup_{\|x\|=1} |\langle Ax, x \rangle|$

" $\geq$ " is clear by Schwartz    " $\leq$ " later

$\sup |\langle x, Ax \rangle| = \max \{ \sup \langle x, Ax \rangle, \sup \langle x, -Ax \rangle \}$

{ Use "Direct method" in the Calculus of Variations

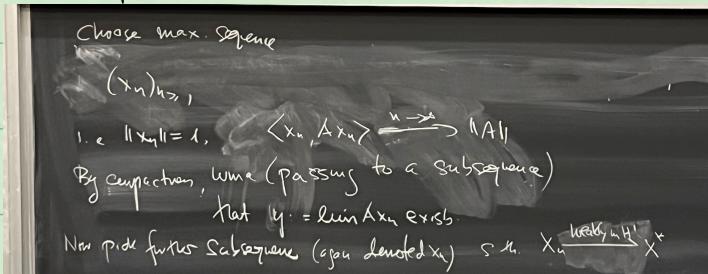
Pick a maximizing sequence  $(x_n)_{n \in \mathbb{N}}$

i.e.  $\|x_n\| = 1$  for all  $n$

$\|Ax_n\| \xrightarrow{n \rightarrow \infty} \|A\|$ .

Since  $A$  is compact, we can choose a convergent subsequence (again denoted  $x_n$ )

By compactness,  $\lim_{n \rightarrow \infty} Ax_n = y \in H$  exists.



Def: For  $f \in L^1(\mathbb{R}^n)$ ,  $\hat{f}(k) : \mathbb{R}^n \rightarrow \mathbb{C}$ :

$$\hat{f}(k) = \int_{\mathbb{R}^n} e^{-2\pi i k \cdot x} f(x) dx, \quad k \cdot x = k_1 x_1 + \dots + k_n x_n$$

By Def:  $\mathcal{F}: f \mapsto \hat{f}$  linear.

$$|\hat{f}(k)| = \left| \int_{\mathbb{R}^n} e^{-2\pi i k \cdot x} f(x) dx \right| \leq \|f\|_1$$

So  $\mathcal{F} : L^1 \rightarrow L^\infty$  bounded.

Goal: Define  $\mathcal{F} : L^2 \rightarrow L^2$  (Note  $C_c^\infty(\mathbb{R}^n) \subseteq L^1 \cap L^2$  dense)

Interaction of  $\mathcal{F}$  with geometric operations.

① translation:  $\forall a \in \mathbb{R}^n$ ,  $T_a : \mathbb{R}^n \rightarrow \mathbb{R}^n$ ,  $T_a(x) = x + a$ .

$$\forall \hat{f} : \mathbb{R}^n \rightarrow \mathbb{C}, \quad (T_a \hat{f})(x) = \hat{f}(T_a^{-1}x) = \hat{f}(x-a)$$

$$\hat{T}_a \hat{f}(k) = \int e^{-2\pi i k \cdot x} \hat{f}(x-a) dx = \int e^{-2\pi i k \cdot y} e^{2\pi i k \cdot a} \hat{f}(y) dy = e^{2\pi i k \cdot a} \hat{f}(k)$$

② derivative:  $\partial_x \hat{f}(k) =$

$$\partial_x \hat{f}(x) = \lim_{h \rightarrow 0} \frac{T_h \hat{f}(x) - \hat{f}(x)}{h}$$

$$\Rightarrow \partial_x \hat{f}(k) = \lim_{h \rightarrow 0} \left( \frac{e^{2\pi i k h} - 1}{h} \right) \hat{f}(k) = 2\pi i k \hat{f}(k)$$

③ dilation:  $\forall \lambda > 0$ ,  $\sigma_\lambda(x) = \lambda x$ ,  $\sigma_\lambda \hat{f}(x) = \hat{f}(\lambda^{-1}x)$

$$\begin{aligned} \hat{\sigma}_\lambda \hat{f}(k) &= \int e^{-2\pi i k \cdot x} \hat{f}(\lambda^{-1}x) dx \quad x = \lambda y, \quad dx = \lambda^n dy \\ &= \int \lambda^n e^{-2\pi i k \cdot \lambda y} \hat{f}(y) dy \\ &= \lambda^n \hat{f}(\lambda k) \end{aligned}$$

④ L-linear transform ( $L \in \mathbb{M}$ )

$$\hat{f} \circ L^{-1}(k) = \int e^{-2\pi i k \cdot (Ly)} f(y) dy = \hat{f}(L^* y)$$

⑤ convolution:

$$\hat{f} * \hat{g}(k) = \int e^{-2\pi i k \cdot x} \int f(x-y) g(y) dy dx = \hat{f}(k) \hat{g}(k) \text{ clear.}$$

Lemma: Let  $g(x) = e^{-\pi|x|^2/\lambda}$  be a gaussian on  $\mathbb{R}^n$

$$\text{then } \hat{g}(k) = \lambda^{-\frac{n}{2}} e^{-\frac{\pi|k|^2}{\lambda}}$$

In particular,  $\hat{g}_1 = \hat{g}_1$

Pf: suffice to compute  $\lambda = 1$

$$\hat{g}_1(k) = \int_{\mathbb{R}^n} e^{-2\pi i k \cdot x} e^{-\pi |x|^2} = \int_{\mathbb{R}^n} e^{-\pi(x+k)(x+k)} dx \cdot e^{-\pi |k|^2}$$

$$\text{Claim: } \int_{-\infty}^{\infty} e^{-\pi(x+k)^2} dx = \int_{-\infty}^{\infty} e^{-\pi x^2} dx = 1.$$

$\hat{g}_1(k)$

↓

$\mathcal{F} = L_1 \rightarrow L_2$  bounded linear. range? complicated.

Will prove:  $\forall f, g \in L^1 \cap L^2$ ,  $\langle \hat{f}, \hat{g} \rangle = \langle f, g \rangle \Rightarrow \|\hat{f}\|_2 = \|f\|_2$

We proved:  $\sigma_x f(x) = f(x'x)$ . For  $f \in L^1$ ,  $\hat{\sigma_x f}(k) = \lambda^n f(\lambda k)$

$$\Rightarrow (\widehat{\lambda^{-\frac{n}{2}} \sigma_x f})(k) = \lambda^{\frac{n}{2}} \hat{f}(\lambda k)$$

$$\|\hat{\sigma_x f}\|_2^2 = \int |\hat{f}(\lambda x)|^2 dx = \lambda^n \|f\|_2^2$$

$$\|\hat{\sigma_x f}\|_2^2 = \int |\lambda^n \hat{f}(\lambda k)|^2 dk = \lambda^n \|f\|_2^2$$

We also compute.  $\hat{g}_1(x) = e^{-\pi|x|^2}$ ,  $\hat{g}_1(x) = \lambda^{-\frac{n}{2}} e^{-\pi|x|^2/\lambda}$

In particular,  $\hat{g}_1 = g_1$

$$\text{Claim: } \lim_{\lambda \rightarrow 0} \hat{g}_\lambda * \psi = \psi$$

$$\lim_{\lambda \rightarrow 0} \hat{g}_\lambda * \psi = \psi$$

Thm: If  $f \in L^1 \cap L^2(\mathbb{R}^n)$ , then  $\|f\|_2 = \|\hat{f}\|_2$

$$\text{Pf: } \iint \hat{g}_\lambda(k) e^{-\pi x \cdot k} e^{2\pi i k \cdot x} f(x) e^{2\pi i k \cdot y} \bar{f(y)} dx dy dk$$

$$= \iint \hat{g}_\lambda(x-y) f(x) \bar{f(y)} dx dy$$

$$= \int (\hat{g}_\lambda * f)(x) \bar{f(x)} dx \xrightarrow{\lambda \rightarrow 0} \|f\|_2^2.$$

$$\iint \hat{g}_\lambda(k) e^{-\pi x \cdot k} e^{2\pi i k \cdot x} f(x) e^{2\pi i k \cdot y} \bar{f(y)} dx dy dk = \int e^{-\pi x \cdot k} \hat{f}(k) dk \xrightarrow[\lambda \rightarrow 0]{\text{MCIT}} \int f(k) dk \quad \checkmark$$

For  $f \in L^2$ , we define  $\mathcal{G}f$  by: pick any  $f_i \rightarrow f$  in  $L^1 \cap L^2$ ,

$$\hat{F} = \lim_{i \rightarrow \infty} \hat{f}_i$$

If  $f$  is carrying in  $L^2 \Rightarrow \hat{f}$  carrying  $\Rightarrow \hat{f} \in L^2$  exists.

Example:  $\hat{f}(x) = \frac{1}{1+|x|} \in L^2 \quad \int |\hat{f}(x)|^2 dx = \int \frac{1}{(1+|x|)^2} dx < \infty$

$$\hat{f}_1 = \chi_{[-1,1]} \hat{f} = \int_{-1}^1 e^{2\pi i k x} \frac{1}{1+|x|} dx$$

For  $f, h \in L^1 \cap L^2$ ,

$$\begin{aligned} \langle \hat{f}, h \rangle &= \iint e^{2\pi i k x} f(x) \overline{h(k)} dk dx \\ &= \int f(x) \int \overline{h(k)} e^{2\pi i k x} dk dx = \langle f, \tilde{h} \rangle, \end{aligned}$$

where  $\tilde{h}(x) = \int e^{2\pi i k x} h(k) dk$ .

So  $\tilde{f}(k) = \tilde{h}$ ,  $\tilde{h}(x) = h(-x)$

Prop:  $\forall f \in L^2(\mathbb{R}^n), \hat{\hat{f}} = \hat{f} = f$

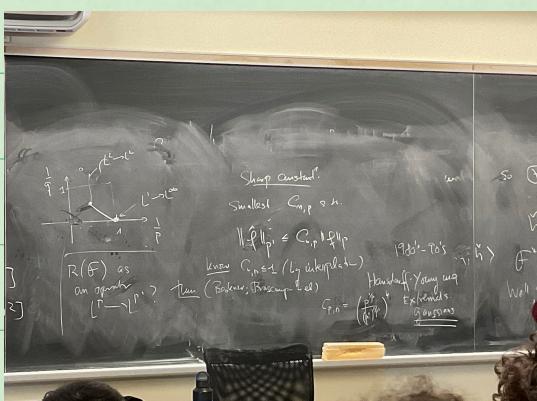
Pf: Consider  $\hat{g}_x * f(x) = \int \hat{g}_x(x-y) \hat{f}(y) dy$

$$\begin{aligned} &= \iint e^{-2\pi i k(x-y)} e^{-2\pi i k y} \hat{f}(y) dy dk \\ &= \int e^{-2\pi i k x} \underbrace{e^{-2\pi i |k|^2} \hat{f}(k)}_{\text{red}} dk \xrightarrow{x \rightarrow 0} \hat{\hat{f}} \end{aligned}$$

Conclusion:  $\mathcal{F}: L^2 \rightarrow L^2$  unitary

$$\begin{aligned} \| \mathcal{F} \|_{L^\infty} &\leq 1 \\ \| \mathcal{F} \|_{L^2 \rightarrow L^2} &= 1 \end{aligned}$$

$\mathcal{F}: L^p \rightarrow L^p \quad \forall p \in [1, 2], \quad \| \mathcal{F} \|_p \leq 1$



Observe:  $\hat{f}$  has compact support  $\Rightarrow \hat{f}$  smooth.

$$\hat{f}(k) = \int_{\mathbb{R}^n} e^{-2\pi i k \cdot x} f(x) dx.$$

$$f \in C_c^\infty, \quad \widehat{\partial_x f} = \int e^{-2\pi i k \cdot x} \partial_x f(x) dx = \int_{2\pi i k_j} e^{-2\pi i k \cdot x} f(x) dx = (2\pi i k_j) \hat{f}(k)$$

Def: Schwartz Space:

$$S := \left\{ f \in \mathbb{R}^n \rightarrow \mathbb{C} \mid x^\alpha \partial_x^\beta f \text{ bounded} \right\}, \quad \begin{matrix} \alpha = (\alpha_1, \dots, \alpha_n) \\ \beta = (\beta_1, \dots, \beta_n) \end{matrix}$$

$$T^\alpha f = \prod (\partial x_j)^{\alpha_j} f, \quad x^\beta = \prod (x_j)^{\beta_j}$$

Fact:  $\mathcal{F}: S \rightarrow S$  continuous by.

Application: Consider Poisson Problem:  $-\Delta u = f$ ,  $\Delta u = \sum \partial_x^2 u$

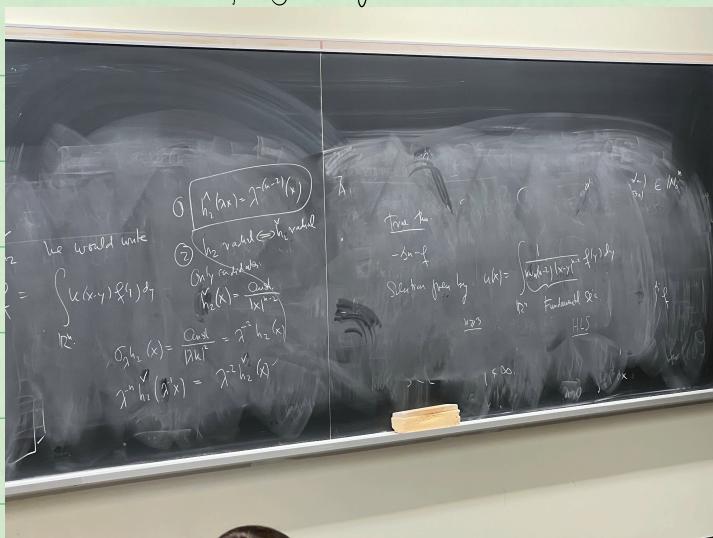
$$-\hat{\Delta} \hat{u}(k) = (2\pi)^2 |k|^2 \hat{u}(k)$$

$$\Rightarrow (2\pi)^2 |k|^2 \hat{u}(k) = \hat{f}(k)$$

$$\Rightarrow \hat{u}(k) = \frac{1}{(2\pi)^2 |k|^2} \hat{f}(k)$$

$$\text{Let } h_2(x) = \frac{1}{(2\pi)^2 |k|^2}$$

$$u(x) = h_2 * f = \int u(x-y) f(y) dy$$



Distributions: (22 Chapter 6, Folland 8, 9).

Distributions: generalized fns.

In  $\mathbb{R}^n$ ,  $C_c^\infty \subseteq S \subseteq L^2 \subseteq L_{loc}^1 \subseteq S' \subseteq \mathcal{D}$  distributions.  
 Shwartz

Let  $\emptyset \neq \Omega \subseteq \mathbb{R}^n$  open connected.

$$\mathcal{D} = C_c^\infty = \{ \psi: \Omega \rightarrow \mathbb{C} \mid \psi \text{ smooth, } \text{supp } \psi \subseteq \Omega \}$$

Elements of  $\mathcal{D}$  are called test functions.

Topology:  $\Psi_n \rightarrow \Psi$  in  $\mathcal{D}$  if =

$$\text{① } \exists K \subseteq \Omega, \text{ s.t. } \forall n, \text{ supp } \Psi_n \subseteq K.$$

$$\text{② } D^\alpha \Psi_n \rightarrow D^\alpha \Psi \quad \forall \text{ multi-index } \alpha.$$

Def: A multi-index is a vector  $\alpha = (\alpha_1, \dots, \alpha_n)$

of  $>0$  integers.  $n = \text{length } \alpha$ .  $|\alpha| = \sum \alpha_i$  order

For  $x \in \mathbb{C}^n$ ,  $x^\alpha = \pi x_1^{\alpha_1} \cdots x_n^{\alpha_n}$ ,  $D^\alpha f = \sum (\partial x_i)^{\alpha_i} f$

$$\mathcal{D}' = \{ T: \mathcal{D} \rightarrow \mathbb{C} \text{ linear cont.} \}.$$

Notation:  $\langle T, \Psi \rangle = T(\Psi)$

i.e. if  $\Psi_n \rightarrow \Psi$  in  $\mathcal{D}$ , then  $T(\Psi_n) \rightarrow T(\Psi)$

Eg: • Let  $f \in L_{loc}^1(\Omega)$

Define  $\langle T, \Psi \rangle = \int_\Omega f(x) \Psi(x) dx$

$$|\langle T, \Psi_n \rangle - \langle T, \Psi \rangle| = \left| \int_\Omega f(x) (\Psi_n - \Psi)(x) dx \right| \leq \sup_{x \in K} |\Psi_n - \Psi| \int_\Omega |f|$$

$$\bullet T_n(\Psi) = \int \Psi$$

$$\bullet S_\alpha(\Psi) := \Psi(\alpha)$$

Let  $T \in \mathcal{D}'$ ,  $\Psi: \mathbb{R}^n \rightarrow \mathbb{C}$

• Define  $\Psi T$  by:

$$\langle \psi T, f \rangle := \langle T, \psi f \rangle$$

• Let  $L \in GL(n, \mathbb{R})$  ( $\det L \neq 0$ )

$$\langle Tf_L, \psi \rangle = \int f(Lx) \psi(x) dx = \int f(y) \psi(L^{-1}y) \frac{1}{|det L|} dy = \langle Tf, (\psi \circ L) \frac{1}{|det L|} \rangle$$

Define  $T \circ L$  by:  $\langle T \circ L, \psi \rangle = \langle T, \frac{1}{|det L|} \psi \circ L \rangle$

How to define  $\hat{T}$ ?

$$\begin{aligned} \langle T\hat{f}, \psi \rangle &= \int \hat{f}(x) \psi(x) dx \\ &= \int f(x) \hat{\psi}(x) dx \end{aligned}$$

$$\text{So } \langle \hat{T}, \psi \rangle = \langle T, \hat{\psi} \rangle$$

• Derivative:  $T \in \mathcal{D}'$ ,  $\alpha = (\alpha_1, \dots, \alpha_n)$  multi-index.

$$\langle T \alpha f, \psi \rangle = \int \partial_x^\alpha f \psi = - \int f \partial_x^\alpha \psi \quad (\text{Integration by part})$$

$$\text{So } \langle D^\alpha T, \psi \rangle = (-1)^{|\alpha|} \langle T, D^\alpha \psi \rangle$$

• Let  $\psi$  be a test fn, define  $f'$  by  $f' = -\psi'(x)$

$$\begin{aligned} \langle f', \psi \rangle &= -\langle f, \psi' \rangle = - \int_{\mathbb{R}} |x| f'(x) \psi(x) dx \\ &= - \int_0^\infty x \psi(x) + \int_{-\infty}^0 x \psi(x) \end{aligned}$$

$$\begin{aligned} \text{IBP} &= -x \psi(x) \Big|_0^\infty + \int_0^\infty \psi(x) dx + x \psi(x) \Big|_{-\infty}^0 - \int_{-\infty}^0 \psi(x) dx \\ &= \int_{-\infty}^\infty \text{sgn} x \psi(x) dx \end{aligned}$$

$$\begin{aligned} \langle f'', \psi \rangle &= \langle \text{sgn} x f', \psi \rangle = - \int_{\mathbb{R}} \text{sgn} x f'(x) \psi(x) dx \\ &= \int_0^\infty \psi' + \int_{-\infty}^0 \psi' \\ &= -\psi \Big|_0^\infty + \psi \Big|_{-\infty}^0 \\ &= 2\psi(0) \end{aligned}$$

$$\text{We define } \langle D^\alpha T, \psi \rangle = (-1)^{|\alpha|} \langle T, D^\alpha \psi \rangle$$

Linear / Continuity: say  $\psi_k \rightarrow \psi$  in  $\mathcal{D}$

$$\forall \beta. \quad D^\beta (\psi_k \rightarrow \psi) \rightarrow 0$$

$$\text{Then } \langle D^\alpha T, \psi_n \rangle = (-1)^{|\alpha|} \langle T, D^\alpha \psi_n \rangle = T(D^\alpha \psi_n)$$

Thm: Let  $T \in \mathcal{D}'$  distribution. Assume  $T$  is positive, i.e.

$$\langle T, \psi \rangle \geq 0, \forall \psi \in \mathcal{D} \text{ with } \psi \geq 0$$

Then  $T$  "is" a measure.  $\exists \mu$  s.t.  $\langle T, \psi \rangle = \int \psi(x) d\mu$ .

$$\mu(A) = \sup \left\{ \int T(\psi) \mid \psi \in \mathcal{D}, \text{ supp } \psi \subset A \right\}.$$

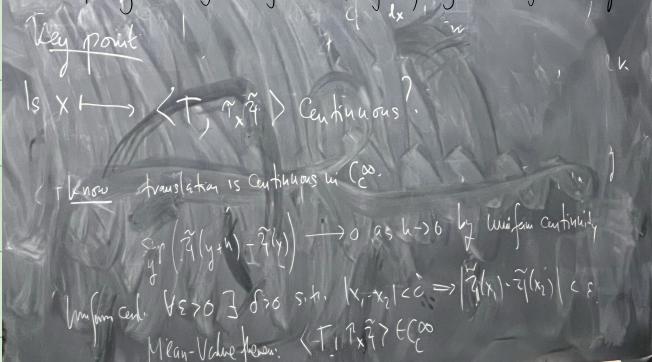
Pf: Define  $\psi * T$  by  $\langle \psi * T, \psi \rangle = \langle T, \psi * \psi \rangle$ , where  $\tilde{\psi}(x) = \psi(x)$

$$\begin{aligned} \langle T * \psi, \psi \rangle &= \int_0^1 \psi(x-y) f(y) dy \psi w dx \\ &= \int \langle T, \tilde{\psi}_x \tilde{\psi} \rangle \psi w dx \end{aligned}$$

Write  $\psi * \tilde{\psi}(x) = \int \psi(y) \tilde{\psi}(y-x) dy$ . approximate by Riemann sum

$$\sum_{j=0}^n \psi(t_j) \tilde{\psi}(t_j - x) |t_j - t_{j-1}|$$

$$\Rightarrow T \left( \sum_{j=0}^n \psi(t_j) \tilde{\psi}(t_j - x) |t_j - t_{j-1}| \right) = \sum_{j=0}^n T(\tau_{t_j} \tilde{\psi}) \psi(t_j)$$



## Def: (Sobolev Spaces)

For  $k \in \mathbb{N} \cup \{0\}$ ,  $p \in [1, \infty]$ , define

$$W^{k,p}(\Omega) = \left\{ T \in \mathcal{D}'(\Omega) \mid T \text{ is represented by a fn } f \in L^p, \right.$$

$\forall |\alpha| \leq k$  multi-index,  $D^\alpha T$  is

represented by a fn  $f_\alpha \in L^p \right\} \subseteq L^p(\Omega)$

$$\Omega = \mathbb{R}^n, P = 2:$$

$$W^{1,2} = H^1 = \left\{ f \in L^2 \mid (1 + |k|^2)^{\frac{1}{2}} f \in L^2 \right\}$$

Norm on  $W^{k,p}$ :  $\sum_{|k| \leq k} \|g_k\|_p = \|\vec{f}\|_{W^{k,p}}$

Example:  $W^{1,p}(\Omega) = \{f \in L^p(\Omega) \mid \exists g_1, \dots, g_n \in L^p(\Omega), \text{ s.t. } \int_\Omega f \partial x_j \psi = \int_\Omega g_j \psi \in C_c^\infty(\Omega)\}$

i.e.  $g_j = \partial_j f$  in sense of distributions

$$\|\vec{f}\|_{W^{1,p}} = \|f\|_{L^p} + \sum_{j=1}^n \|g_j\|_{L^p}$$

Prop:  $W^{1,p}(\Omega)$  is complete.

Pf: Let  $\{f_n\}$  be a Cauchy sequence in  $W^{1,p}$

i.e.  $\{f_n\}$  is Cauchy in  $L^p$ , also the distributional derivatives  $\{(g_j)_k\}$ .

Completeness of  $L^p \Rightarrow f, g_1, \dots, g_n \in L^p$ ,

$f_k \rightarrow f, (g_j)_k \rightarrow g_j$ . Now  $g_j = \partial_j f$  as distributions.

Let  $\psi$  be a test function

$$\begin{aligned} \langle T_{\partial_j} f, \psi \rangle &= - \int f(x) \partial_j \psi(x) dx \\ &= - \lim_{k \rightarrow \infty} \int \overset{f \in W^{1,p}}{f_k(x)} \partial_j \psi(x) dx \\ &= \lim_{k \rightarrow \infty} \int (g_j)_k(x) \psi(x) dx \\ &= \int g_j(x) \psi(x) dx \end{aligned}$$

Sobolev Inequalities:

Consider  $W^{1,p}(\mathbb{R}^n)$ ,  $1 \leq p < n$ , set  $p^* = \frac{np}{n-p}$

$$\exists C = C(n, p), \text{ s.t. } \|u\|_{p^*} \leq C(n, p) \|\nabla u\|_p \quad = \sum_{j=1}^n \|\partial_j u\|_p$$

$$\text{LHS} = \left( \int |u|^{p^*} dx \right)^{\frac{1}{p^*}} \sim m^{\frac{n}{p^*}}$$

$$\text{RHS} = \left( \int |\nabla u|^p dx \right)^{\frac{1}{p}} \sim m^{\frac{n-p}{p}} \quad \Rightarrow \quad \frac{n}{p^*} = \frac{n-p}{p}$$

Thm: (Hahn-Banach):

Let  $X$  be a  $\mathbb{R}$ -vector space,  $p$  a sublinear functional on  $X$ ,  $M \subseteq X$  a subspace,  $f$  a linear functional on  $M$  s.t.  $f(x) \leq p(x)$ ,  $\forall x \in M$ .

Then  $\exists F: X \rightarrow \mathbb{R}$  linear s.t.  $F(x) \leq p(x)$ ,  $\forall x \in X$ ,  $F|_M = f$

Cor:  $X$  a normed vector space,

If  $M \subseteq X$ ,  $f: M \rightarrow \mathbb{R}$  bounded linear, then

$\exists$  a bdd linear  $F: X \rightarrow \mathbb{R}$ , s.t.  $\|F\|_X = \|f\|_M$

$$\text{Def: } P(x) := \|f\|_M \|x\|$$

$$\text{then } F(x) \leq \|f\|_M \|x\|$$

Remark:  $X$  is a locally convex topological V.S. <sup>TVS</sup>

$\Rightarrow$  cor holds for cont. extension

e.g.  $L^p[0,1]$  when  $0 < p \leq 1$ , there are no cont. linear functionals

on  $L^p$ , the only open convex sets are  $\{0\}$  and  $L^p$

Convexity:  $X$  a TVS

Def: The gauge function of  $K \subseteq X$  with  $0 \in K$  is

$$P_K(x) = \inf \{r > 0 : x \in rK\}.$$

E.g.:  $X$  has a norm  $\|\cdot\|$ ,  $K = \{x : \|x\| \leq 1\}$ .  $P_K(x) = \|x\|$

•  $X$  is  $\mathbb{R}^n$  with  $\|\cdot\|_p$ ,  $K = \{x : \|x\|_p \leq 1\}$ ,  $P_K(x) = \|x\|_p$

•  $f: X \rightarrow \mathbb{R}$ ,  $K = \{x : |f(x)| \leq 1\}$ .  $P_K(x)$  almost a norm

Assume  $X$  is a NVS

Lemma:  $C \subseteq X$  open, convex, and  $0 \in C$ ,  $P_C = P_C$ .

①  $P$  is sublinear.

②  $\exists M > 0$ , s.t.  $P(x) \leq M \cdot \|x\|$ ,  $\forall x$ .

③  $C = \{x : P(x) = 1\}$ .

Def: ②  $\Leftrightarrow \exists r > 0$ , s.t.  $B_{r(0)} \subseteq C$

$$\Rightarrow P(x) \leq \frac{1}{r} \|x\|, M = \frac{1}{r}$$

③ If  $x \in C \Rightarrow \exists \varepsilon > 0$ , s.t.  $(1+\varepsilon)x \in C$ .

$$\Rightarrow P(x) \leq \frac{1}{1+\varepsilon} < 1$$

If  $P(x) < 1 \Rightarrow \exists r < 1$  s.t.  $\frac{1}{r}x \in C$ .

Convexity:  $r(\frac{1}{r}x) + (1-r)0 = x \in C$ .

① If  $\lambda > 0$ ,

$$\begin{aligned} P(\lambda x) &= \inf \{r > 0 \mid \lambda x \in B_r\} \\ &= \lambda \inf \{\lambda r > 0 \mid x \in B_r\} \\ &= \lambda P(x) \end{aligned}$$

$$P(x+y) \leq P(x) + P(y) + \varepsilon, \quad \forall \varepsilon > 0.$$

Remark: If  $C$  is open, convex, symmetric ( $-c \in C$ )

and bounded w.r.t.  $\|\cdot\|$

$\Rightarrow P$  is a norm equivalent to  $\|\cdot\|$

Def: A Hyperplane  $H \subseteq X$  is

$H = \{x : f(x) = \alpha\}$  for some linear functional  $f$  and  $\alpha \in \mathbb{R}$ .

Lemma: (Brezis)

$H = \{f = \alpha\}$  is closed iff  $f$  is continuous.

Def  $A, B \subseteq X$ ,  $H = \{f = \alpha\}$

$H$  separates  $A$  and  $B$  if  $f(a) \leq \alpha < f(b)$   $\forall a \in A, b \in B$ .

$H$  strictly separates  $A$  and  $B$  if  $\exists \varepsilon > 0$ ,  $f(a) + \varepsilon \leq \alpha \leq f(b) - \varepsilon$ .

Lemma: If  $C \subseteq X$  open convex,  $x_0 \notin C$ , then  $\exists f = f(x_0)$  separates  $x_0$  and  $C$ .

Pf: Assume  $0 \in C$ .  $M = \text{span}\{x_0\}$ .

$f: M \rightarrow \mathbb{R}$ ,  $f(tx_0) = t$ ,  $P$  is the gauge function of  $C$ .  
 $|f(x_0)| \leq P(x_0)$  Lemma: since  $C = \{p(x) < 1\}$ .

$\forall t \in \mathbb{R}$ ,  $f(tx_0) = t f(x_0) \leq t P(x_0) = P(tx_0)$ .

$\Rightarrow f$  dominated by  $M$ .

$\Rightarrow$  An ext<sup>n</sup>  $F: X \rightarrow \mathbb{R}$ ,  $\|F\|_X = \|P\|_M$ .

$x \in C$   $F(x) \leq P(x) < 1$

$F(x_0) = 1$

$H = \{F = 1\}$

Key idea: separating  $A$  from  $B \Leftrightarrow$  separating  $A-B$  from  $0$ .

Geometric H-B for open sets:

$A, B$  disjoint, non-empty convex sets and  $A$  is open

$\Rightarrow \exists$  a separating hyperplane.

Pf:  $C = A - B$ ,  $x_0 = 0$

Geometric H-B, compact

$A, B$  disjoint convex

$A$  is closed  $\Leftrightarrow B$  is compact

$\Rightarrow \exists$  a strictly separating hyperplane.

RF:  $C = \bar{A} - \bar{B}$  closed, convex,  $0 \notin C$

$\Rightarrow \exists B_{r(0)}$  disjoint from  $C$

$\Rightarrow \exists f = \varphi$  separates  $B_{r(0)} \cup C$

i.e.,  $f(a-b) \leq \varphi \leq f(v)$ ,  $\forall a \in A, b \in B, v \in B_{r(0)}$

$f(a-b) \leq -r \cdot \|f\|$  (Take inf of  $v$ )

$$\Rightarrow f(a) + \frac{r\|f\|}{2} \leq f(b) - \frac{r\|f\|}{2}$$

$$\text{pick } \beta = f(a) + \frac{r\|f\|}{2} \leq \beta \leq f(b) - \frac{r\|f\|}{2}$$

$\Rightarrow$  strictly separate.

$A, B$  disjoint convex  $\neq \emptyset$

finite dim  $\Rightarrow$  always separate

$A$  open  $\Rightarrow$  separate

$A$  closed,  $B$  compact  $\Rightarrow$  strictly separate

$A, B$  closed  $\Rightarrow$  might not separate.

Recall:  $C$  open, convex,  $0 \in C$ ,  $\mathcal{P}$  gauge

$$\Rightarrow C = \{x \mid p(x) < 1\}$$

$$\text{Core } \bar{C} = \{x \mid p(x) \leq 1\}$$

$\Psi: X \rightarrow [0, +\infty]$  'nice' convex fn

$$K = \{x \mid \Psi(x) \leq 1\} \text{ closed convex}$$

$$K = \{x \mid P_K(x) \leq 1\}$$

For  $\lambda > 0$

$$\{x \mid \Psi(x) \leq \lambda\} = \{x \mid P_K(x) \leq \lambda\}$$

Convex fn's  $\Psi: X \rightarrow (-\infty, +\infty]$

Claim:  $\{x | \Psi(x) \leq \lambda\}, \forall \lambda \in \mathbb{R}$

are closed  $\Leftrightarrow \Psi$  is lower semi-continuous (l.s.c)

if  $x_n \rightarrow x, \liminf \Psi(x_n) \geq \Psi(x)$

Claim:  $\Psi$  is convex  $\Rightarrow \{x | \Psi(x) \leq \lambda\}$  is convex.

Def:  $\text{epi}(\Psi) = \{(x, \lambda) | \Psi(x) \leq \lambda\} \subseteq X \times \mathbb{R}$ .

Claim: ①  $\Psi$  is l.s.c  $\Leftrightarrow \text{epi}(\Psi)$  closed

②  $\Psi$  is convex  $\Leftrightarrow \text{epi}(\Psi)$  convex

③  $\Psi$  is proper  $\Leftrightarrow \text{epi}(\Psi) \neq \emptyset$  ( $\Psi \neq \infty$ )

H-B = P is gauge, want  $f \in X^*, f \leq P$

Def:  $P^*: X \rightarrow (-\infty, \infty]$  indicator fn

$$P^*(f) = \begin{cases} 0 & f \leq P \\ +\infty & \text{otherwise.} \end{cases}$$

Def:  $\Psi$  is proper  $\Rightarrow \Psi^*(f) = \sup_x f(x) - \Psi(x)$

<u>Examples.</u>
1. P is gauge fn. of some C, $\text{dom}+C$
$P^*(f) := \sup_x f(x) - P(x)$
case 1. $f \leq P \Rightarrow P^*(f) = 0$ $f(x) - P(x) \leq 0$
case 2. $f(x_0) > P(x_0)$
$P^*(f) \geq \sup_{\lambda > 0} f(\lambda x_0) - P(\lambda x_0) = +\infty$

$\mathcal{D} = \Psi: \mathbb{R} \rightarrow \mathbb{R}, P > 1$

$$\Psi(a) = \frac{1}{P} a^P, \text{ identifying } \mathbb{R}^* = \mathbb{R}$$

$$\text{If } b \in \mathbb{R}, \Psi^*(b) = \inf_a ab - \frac{1}{P} a^P = \frac{1}{P} b^P$$

Young's inequality:  $\Psi: X \rightarrow (-\infty, +\infty], \forall x \in X, \exists x^*$

$$f(x) \leq \Psi(x) + \Psi^*(x)$$

Prop: if  $\Psi: X \rightarrow (-\infty, +\infty]$  is convex, l.s.c., proper

$\Rightarrow$  so is  $\Psi^*$

Pf:

$$\begin{aligned} \text{Proof:} \quad & \text{epi}(\Psi^*) = \left\{ (f, \lambda) : \lambda \geq \sup_{x \in X} (f(x) - \Psi(x)) \right\} \\ &= \bigcap_{x \in X} \left\{ (f, \lambda) : \lambda \geq f(x) - \underbrace{\Psi(x)}_{\substack{\text{convex} \\ \text{on } X}} \right\} \\ \Rightarrow \text{epi}(\Psi^*) &= \bigcap \text{closed + convex} \\ &= \text{closed + convex} \end{aligned}$$

Want  $\text{epi}(\Psi^*) \neq \emptyset$

Need  $f \in X^*$  and  $K \in \mathbb{R}$  s.t.  $f(x) - \Psi(x) \leq K \quad \forall x \in X$   
rearrange:  $\Psi(x) \geq f(x) - K \quad \forall x \in X$

Choose  $x_0 \in X$  s.t.  $\Psi(x_0) < +\infty$

and  $\lambda_0 \in \mathbb{R}$  s.t.  $\lambda_0 \Psi(x_0) \Rightarrow (x_0, \lambda_0) \in \text{epi}(\Psi)$

$\Rightarrow \exists$  a strictly sep H:  $[F = \infty] \subseteq X \times \mathbb{R}$

$\exists f \in X^*$  and  $t \in \mathbb{R}$  s.t.  $F(x, \lambda) = f(x) + t\lambda$ .

If  $\Psi(x) < +\infty$

$$f(x_0) + t\lambda_0 < \alpha < f(x) + t\Psi(x)$$

Set  $x = x_0$

$$f(x_0) + t\lambda_0 < f(x_0) + t\Psi(x_0)$$

$$\Rightarrow t(\Psi(x_0) - \lambda_0) > 0 \Rightarrow t > 0$$

$$\text{so } f(x) + t\Psi(x) > \alpha \quad \forall x \in X$$

$$\Psi(x) > \left(-\frac{f}{t}\right)(x) + \frac{\alpha}{t}$$

$$\Psi^*\left(-\frac{f}{t}\right) < \frac{-\alpha}{t} < +\infty$$

Example: Fix  $\Psi: X \rightarrow (-\infty, +\infty]$ .

1. Translate by constant  $\alpha \in \mathbb{R}$

$$(\Psi + \alpha)(x) = \Psi(x) + \alpha.$$

$$(\Psi + \alpha)^*(x) = \sup_{f \in \mathcal{F}} f(x) - \Psi^*(x) - \alpha$$

$$= \Psi^*(f) - \alpha.$$

2. Scale:  $\lambda \in \mathbb{R}$   $(\lambda \Psi)(x) = \lambda \Psi(x)$

$$(\lambda \Psi)^*(x) = \sup_{f \in \mathcal{F}} f(x) - \lambda \Psi^*(x)$$

$$= \lambda \sup_{f \in \mathcal{F}} f - \lambda \Psi^*(x)$$

$$= \lambda \Psi^*(\lambda f)$$

3.  $(\Psi + g)$ ,  $g \in X^*$

$$(\Psi + g)^*(f) = \sup_{f \in \mathcal{F}} f - \Psi^*(f) - g$$

$$= \Psi^*(f) - g$$

Claim:  $\Psi \leq \Psi^* \Rightarrow \Psi^* \geq \Psi^*$

Avoid  $X \neq X^{**}$

Def:  $\Psi: X \rightarrow (-\infty, +\infty]$

$$\Psi^{**}: X \rightarrow (-\infty, +\infty]$$

$$\Psi^{**}(x) = \sup_{f \in \mathcal{F}} f(x) - \Psi^*(f)$$

Thm: (Fenchel-Moreau):

If  $\Psi$  convex, proper, l.l.c.

then  $\Psi = \Psi^{**}$

Pf: Case 1:  $\Psi > 0$

$$f(x) - \Psi^*(f) \leq \Psi(x), \forall f, x$$

$$\Rightarrow \Psi^{**}(x) \leq \Psi(x)$$

Say  $\psi^{**}(x_0) < \psi(x_0)$

Strictly sep.  $(x_0, \psi^{**}(x_0))$  and  $\text{epi}(\psi)$

$\exists f \in X^*, t \in \mathbb{R}$

$$f(x_0) + t\psi^{**}(x_0) < f(x) + t\lambda \quad \forall (x, \lambda) \in \text{epi}(\psi)$$

Fix some  $x$ , s.t.  $\psi(x) < +\infty$

Let  $\lambda \rightarrow +\infty, \Rightarrow t \geq 0$ .

$$\begin{aligned} & \text{strictly sep } (x_0, \psi^{**}(x_0)) \text{ and } \text{epi}(\psi) \text{ by } [f = \infty] \\ & \exists f \in X^*, t \in \mathbb{R} \\ & f(x_0) + t\psi^{**}(x_0) < f(x) + t\lambda \\ & \forall (x, \lambda) \in \text{epi}(\psi) \\ & f(x) + t\psi(x) \geq \infty \quad \forall x \text{ since } t \geq 0 \\ & f(x) + (t+\varepsilon)\psi(x) \geq \infty \\ & \text{Fix some } x \text{ s.t. } \psi(x) < +\infty \quad \text{since } \psi \geq 0. \\ & \text{Let } \lambda \rightarrow +\infty \\ & \Rightarrow t \geq 0 \\ & \left( \frac{-f}{t+\varepsilon} \right)(x) - \psi(x) \leq \frac{-\infty}{t+\varepsilon} \\ & \psi^* \left( \frac{-f}{t+\varepsilon} \right) \leq \frac{-\infty}{t+\varepsilon} \end{aligned}$$

Case 2:

Case 2.  $\psi$  is arbitrary

Choose  $f_0 \in X^*$  s.t.  $\psi^*(f_0) < +\infty$

$\bar{\psi}(x) := \psi(x) + \psi^*(f_0) - f_0(x) \geq 0$  Young

$(\bar{\psi})^* = \bar{\psi}$  by case 1.

$\bar{\psi} = \psi - f_0 + \psi^*(f_0)$

$(\bar{\psi})^*(g) = \psi^*(f_0 + g) - \psi^*(f_0)$

Theorem (Fenchel-Moreau)

If  $\psi$  is convex, proper, l.s.c., then  $\psi = \psi^{**}$ .

Proof case 2 contd

$$\begin{aligned} \bar{\psi}(x) &:= \sup_g g(x) - \psi^*(g) \\ &= \sup_g g(x) - \bar{\psi}^*(f_0 + g) + \psi^*(f_0) \end{aligned}$$

$$\begin{aligned} &= \psi^*(f_0) + \sup_g g(x) - \psi^*(f_0 + g) \\ &= \psi^*(f_0) + f_0(x) + \sup_g (g + f_0)(x) \\ &\stackrel{?}{=} (\psi^*(f_0) + g) \\ &= \psi^*(f_0) - f_0(x) + \psi^*(x) \\ &= \bar{\psi}^*(x) = \psi^*(f_0) - f_0(x) + \psi^{**}(x) \\ &= \bar{\psi}(x) = (\psi^*(f_0) - f_0(x)) + \psi(x) \\ &\Rightarrow \psi^{**} = \psi \end{aligned}$$

□