

## MATH 105 HW 8

1. Done.

2. Let  $(f_1, f_2, \dots, f_n, \dots)$  be a sequence of measurable functions that converge almost everywhere to  $f$  as  $k \rightarrow \infty$ .

(a) Define  $X(k, \epsilon) = \{x \in I^d : \forall n \geq k, |f_n(x) - f(x)| < \frac{\epsilon}{2}\}$  where  $d$  is the dimension. Fix  $\epsilon > 0$ . Since  $f_n(x) \rightarrow f(x)$  for almost every  $x$ , we have:  $\bigcup_k X(k, \epsilon) \cup Z(\epsilon) = I^d$ , where  $Z(\epsilon)$  is a zero set. [d refers to box in question]

Let  $\epsilon > 0$  be given. By continuity of measure,  $\lim_{k \rightarrow \infty} m(X(k, \epsilon)) = \prod (b_i - a_i)$ . We can choose indices  $k_1, k_2, \dots$  s.t.  $X_k = X(k_k, \epsilon)$  and  $m(X_k^c) < \frac{\epsilon}{2^k}$ .

Let  $X = \bigcap_k X_k$ . Then  $m(X^c) < \epsilon$ .  
(since  $m(X^c) \leq \sum_{k=1}^{\infty} m(X_k^c) = \epsilon$ ).

Claim:  $f_n$  converges uniformly on  $X$ . Given  $\delta > 0$ , choose and fix  $k$  s.t.  $\frac{1}{k} < \delta$ . Then, for all  $n \geq k$ , we have:  $x \in X \Rightarrow x \in X_k = X(k, \epsilon) \Rightarrow |f_n(x) - f(x)| < \frac{1}{k} < \delta$ .

Hence,  $f_n$  converges uniformly to  $f$  less the  $\epsilon$ -set  $X^c$ .

(b) Egoroff's Theorem remains true. Since the set  $S$  is unbounded but has finite measure, consider balls  $B_r(0)$  where  $r \rightarrow \infty$ .  $\exists M \in \mathbb{N}$  s.t.  $|m(B_M(0)) - m(S)| < \frac{\epsilon}{2}$ . Then, we can repeat the argument in part (a) for the bounded set  $B_M(0)$ , i.e.  $f_n$  converges uniformly to  $f$  less an  $\frac{\epsilon}{2}$ -set  $X^c$ . Then, excluding  $S \setminus B_M(0)$  and the  $\frac{\epsilon}{2}$ -set  $X^c$ , we are done  $\Rightarrow$  Egoroff's Theorem still true.

(c) Consider the moving bump functions  $f_k(x) = \begin{cases} 1 & x \in [k, k+1) \\ 0 & \text{else.} \end{cases}$   
Then,  $(f_n)$  converges ~~to~~ almost everywhere, since  $\forall x, f_{\lfloor x \rfloor}(x) = 1$   
 $f_k(x) = 0 \quad \forall k > \lfloor x \rfloor$ .

Hence  $\lim_{n \rightarrow \infty} f_n(x) = 0$ .  $\therefore (f_n)$  converges everywhere.

However, the sequence of functions is not uniformly convergent on  $\mathbb{R}$ .

Uniform convergence would imply  $\forall \epsilon, \exists N$  s.t.  $n > N, |f_n(x) - f(x)| < \epsilon$ . But  $f = 0$  i.e.  $|f_n(x)| < \epsilon$ . This is false since  $f_n(x) = 1$  for  $x \in [n, n+1)$  hence the series is not uniformly convergent on  $\mathbb{R}$ .

2nd)  $(f_n)_n : \mathbb{R}^n \rightarrow \mathbb{R}$   $\epsilon > 0$ .

Partition  $\mathbb{R}^n$  into cubes. For each cube,  $\forall \epsilon > 0$ ,  $\exists$  a set  $S'$  s.t.  $\bigcup S'$  will enable the sequence of functions  $(f_n)_n$  to converge uniformly on it.

Let  $K \subset \mathbb{R}^n$  be a compact set. Then  $K$  can be ~~expressed~~ bounded by a collection of <sup>finite</sup> cubes since  $K$  is bounded. Let the cubes be  $Q_1, Q_2, \dots, Q_k$ . Then, from each cube  $Q_i$ , there exists a set  $S_i$  s.t.  $\bigcup S_i$  will enable the sequence of functions  $(f_n)_n$  to converge uniformly on it. Hence, let  $S = \bigcup S_i$ . Then  $K \cap S$  will enable the sequence of functions  $(f_n)_n$  to converge uniformly, since we can take max over a finite number of cubes. i.e.  $\forall \epsilon > 0$ ,  $\exists N_1, N_2, \dots, N_k$  s.t.  $n_i > N_i \Rightarrow |f_{n_i}(x) - f(x)| < \epsilon$ . Then, set  $N = \max_{1 \leq i \leq k} N_i \Rightarrow n > N \Rightarrow |f_n(x) - f(x)| < \epsilon$ .

Hence, the functions restricted to  $K \cap S^c$  converges uniformly as desired.

3rd)  $(\mathbb{R}^n, \|\cdot\|_1)$  normed vector space  $\|(x_1, x_2, \dots, x_n)\|_1 = \sum |x_i|$

$T: \mathbb{R}^n \rightarrow \mathbb{R}^n$  be linear operator, given by  $T_{ij}$ , sends  $x_j$  to  $y_i$ .  $y_i = \sum_j T_{ij} x_j$

$$\|T\| := \sup \left\{ \frac{\|Tv\|_1}{\|v\|_1} : v \neq 0 \right\} = \sup \left\{ \|Tv\|_1 : \|v\|_1 = 1 \right\}$$

$$\text{let } y = Tv. \text{ Then } \|y\| = \sum_i |y_i| = \sum_i \left| \sum_j T_{ij} v_j \right|$$

$$\Rightarrow \sup_{v: \|v\|_1=1} \|y\| = \sup_{v: \|v\|_1=1} \sum_j \left| \sum_i T_{ij} v_j \right| = \sup_j \sum_i |T_{ij} v_j| \quad \text{subject to } |v_1| + |v_2| + \dots + |v_n| = 1$$

$$= \sup_j \sum_i |T_{ij}| \quad T = \begin{bmatrix} -t_1^T & - \\ -t_2^T & - \\ \vdots & \vdots \\ -t_n^T & - \end{bmatrix}$$

For me, I think of it geometrically, i.e. the sup is achieved at one of the vertices of the  $\ell_1$  ball, i.e. when  $v = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$  or  $\begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}$  or  $\dots$   $\begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$ . This is because at every step, the  $\sum_i |T_{ij} v_j|$  is just a system of linear equalities. Hence, if we take derivative with respect to any component, it will do no more by shifting that component towards 0 or 1, depending on the sign of the derivative.

Hence, this problem is equivalent to  $\sup \sum_i |T_{ij} v_j|$ , for  $v \in \left\{ \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix} \right\}$   
 $= \sup \|t_i\|_1$ , where  $t_i$  is the  $i$ th row of  $T$

$$\therefore \|T\| = \sup \|t_i\|_1 = \max \|t_i\|_1 \quad (\text{i.e. max } \ell_1 \text{ norm of columns})$$

Another solution is by algebra:

$$\begin{aligned}
 \|Tv\|_1 &= \left\| \begin{bmatrix} | & | & \dots & | \\ t_1 & t_2 & \dots & t_n \\ | & | & \dots & | \\ \vdots & \vdots & \dots & \vdots \\ | & | & \dots & | \\ v_1 & v_2 & \dots & v_n \\ | & | & \dots & | \end{bmatrix} \right\|_1 = \left\| v_1 \begin{bmatrix} | \\ t_1 \\ | \\ \vdots \\ | \end{bmatrix} + v_2 \begin{bmatrix} | \\ t_2 \\ | \\ \vdots \\ | \end{bmatrix} + \dots + v_n \begin{bmatrix} | \\ t_n \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 \\
 &\leq \left\| v_1 \begin{bmatrix} | \\ t_1 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 + \left\| v_2 \begin{bmatrix} | \\ t_2 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 + \dots + \left\| v_n \begin{bmatrix} | \\ t_n \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 \\
 &= |v_1| \left\| \begin{bmatrix} | \\ t_1 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 + |v_2| \left\| \begin{bmatrix} | \\ t_2 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 + \dots + |v_n| \left\| \begin{bmatrix} | \\ t_n \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 \\
 &\leq (|v_1| + |v_2| + \dots + |v_n|) \max \left( \left\| \begin{bmatrix} | \\ t_1 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1, \dots, \left\| \begin{bmatrix} | \\ t_n \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 \right) \\
 &= \|v\|_1 \cdot \max \left( \left\| \begin{bmatrix} | \\ t_1 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1, \dots, \left\| \begin{bmatrix} | \\ t_n \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 \right) \\
 &= \max \left( \left\| \begin{bmatrix} | \\ t_1 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1, \dots, \left\| \begin{bmatrix} | \\ t_n \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 \right).
 \end{aligned}$$

Note that this max is achievable, i.e. when ~~v = sign(t<sub>i</sub>)~~

$$v = t_{\max} \text{ where } t_{\max} = \operatorname{argmax} \left( \left\| \begin{bmatrix} | \\ t_1 \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1, \dots, \left\| \begin{bmatrix} | \\ t_n \\ | \\ \vdots \\ | \end{bmatrix} \right\|_1 \right)$$

hence  $\|T\|_1 = \max \|t_i\|_1$  as desired. (max of  $\ell_1$  norms of columns)

$$\begin{aligned}
 3(b) \quad \|T\|_\infty &= \sup \left\{ \frac{\|Tv\|_\infty}{\|v\|_\infty} : v \neq 0 \right\}, \quad \|v\|_\infty = \max_{1 \leq i \leq n} |v_i|. \\
 &= \sup \left\{ \|Tv\|_\infty : \|v\|_\infty = 1 \right\}
 \end{aligned}$$

$$\|Tv\|_\infty = \left\| \begin{bmatrix} -t_1^T \\ -t_2^T \\ \vdots \\ -t_n^T \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \right\|_\infty = \max \left( |t_1^T v|, |t_2^T v|, \dots, |t_n^T v| \right)$$

Note by Cauchy-Schwarz, actually Hölder's inequality,  $|t_i^T v| \leq \|t_i\|_1 \|v\|_\infty = \|t_i\|_1$

$$\begin{aligned}
 \text{hence } \|T\|_\infty &= \max \left( |t_1^T v|, |t_2^T v|, \dots, |t_n^T v| \right) \\
 &\leq \max \left( \|t_1\|_1, \|t_2\|_1, \dots, \|t_n\|_1 \right)
 \end{aligned}$$

Note that equality is achievable when  $v_i = \operatorname{sign}(t_{ij})$

$$\text{hence, } \|T\|_\infty = \max \|t_i\|_1 = \max \ell_1 \text{ norm of rows.}$$

4. Hölder's inequality: For conjugates  $p, q$  s.t.  $\frac{1}{q} + \frac{1}{p} = 1$ ,  $p, q \geq 1$ .

$$\left( \sum_{i=1}^n |x_i y_i| \right) \leq \left( \sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}} \left( \sum_{i=1}^n |y_i|^q \right)^{\frac{1}{q}}$$

i.e.  $|x^T y| \leq \|x\|_p \|y\|_q$

Firstly, we prove the Young's inequality: (Most parts below are referenced from UPENN MATH browser/courses/2011/Math361/Notes/YMandH.pdf.)

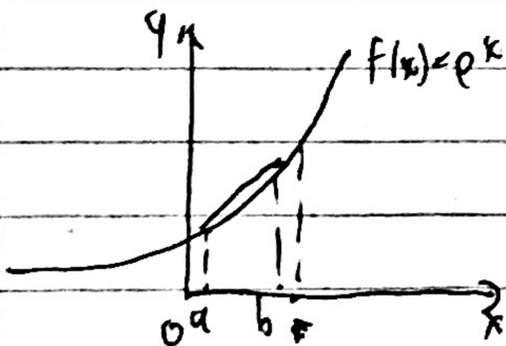
Young's Inequality

When  $1 < p < \infty$ ,  $a, b \geq 0$ ,

Young's inequality gives:  $ab \leq \frac{p-1}{p} a^{\frac{p}{p-1}} + \frac{1}{p} b^p$

Lemma: For  $t \in [0, 1]$ , then  $e^{ta+(1-t)b} \leq te^a + (1-t)e^b$

This is due to the secant line trick in Evan Chen's notes.



Because  $f(x) = e^x$  is convex, the line joining  $(a, e^a)$  and  $(b, e^b)$  lies above the curve  
 $\Rightarrow \forall t \in [0, 1], f(ta + (1-t)b) \leq tf(a) + (1-t)f(b)$

$$\Rightarrow e^{ta+(1-t)b} \leq te^a + (1-t)e^b$$

Proof: [Young] LHS =  $ab = e^{\log a + \log b} = e^{\frac{p-1}{p} \frac{p}{p-1} \log a + (1 - \frac{p-1}{p}) \frac{1}{1 - \frac{p-1}{p}} \log b}$   
 $\leq \frac{p-1}{p} e^{\frac{p}{p-1} \log a} + (1 - \frac{p-1}{p}) e^{\frac{1}{1 - \frac{p-1}{p}} \log b}$   
 $= \frac{p-1}{p} a^{\frac{p}{p-1}} + \frac{1}{p} b^p = \text{RHS. } \boxed{\text{as desired}}$

Hölder's Proof (Hölder's)  $\sum_{i=1}^n x_i y_i \leq \|x\|_p \|y\|_q$

LHS =  $\sum_{i=1}^n \frac{x_i}{\|x\|_p} \frac{y_i}{\|y\|_q} \leq \sum_{i=1}^n \frac{|x_i|}{\|x\|_p} \frac{|y_i|}{\|y\|_q}$  (must as well make both positive to prove a tighter bound)

$$\leq \sum_{i=1}^n \frac{p-1}{p} \frac{|x_i|}{\|x\|_p} \frac{|x_i|}{\|x\|_p} + \frac{1}{p} \frac{|y_i|^p}{\|y\|_q^p}$$

(because by Young's inequality,  $\frac{|x_i|}{\|x\|_p} \cdot \frac{|y_i|}{\|y\|_q} \leq \frac{p-1}{p} \left( \frac{|x_i|}{\|x\|_p} \right)^{\frac{p}{p-1}} + \frac{1}{p} \left( \frac{|y_i|}{\|y\|_q} \right)^p$ )

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(continued)

$$= \sum_{i=1}^n \left( \frac{p-1}{p} \frac{|x_i|^2}{\|x\|_q^2} + \frac{1}{p} \frac{|y_i|^p}{\|y\|_p^p} \right)$$

$$= \frac{p-1}{p} \frac{1}{\|x\|_q^{\frac{p-1}{p-1} \cdot \frac{p}{p-1}}} \sum_{i=1}^n |x_i|^{\frac{p}{p-1}} + \frac{1}{p} \frac{1}{\|y\|_p^p} \sum_{i=1}^n |y_i|^p$$

$$= \frac{p-1}{p} \cdot \frac{\|x\|_q^{\frac{p}{p-1}}}{\|x\|_q^2} + \frac{1}{p} \frac{\|y\|_p^p}{\|y\|_p^p} = 1$$

$\therefore \sum_{i=1}^n \frac{x_i}{\|x\|_q^{\frac{p}{p-1}}} \frac{y_i}{\|y\|_p} \leq 1 \Rightarrow \left[ \sum_{i=1}^n x_i y_i \leq \|x\|_q \|y\|_p \right]$  as desired.

Minkowski Theorem

If  $1 \leq p < \infty$ , whenever  $x, y \in V_p$  we have  $\|x+y\|_p \leq \|x\|_p + \|y\|_p$ .

$$\left( \sum_{i=1}^n |x_i + y_i|^p \right)^{\frac{1}{p}} \leq \left( \sum_{i=1}^n |x_i|^p \right)^{\frac{1}{p}} + \left( \sum_{i=1}^n |y_i|^p \right)^{\frac{1}{p}}$$

I followed [Bath.ac.uk/masmdp/ueg/dir.bho/mtesp54to61.pdf](http://math.ac.uk/masmdp/ueg/dir.bho/mtesp54to61.pdf).

$\|f+g\|_p \leq \|f\|_p + \|g\|_p$ . Suppose  $f+g \geq 0$  pointwise. If  $\|f+g\|_p = 0$ , then  $f+g = 0$  a.e.  
 $\Rightarrow \|f+g\|_p = 0 \leq \|f\|_p + \|g\|_p = 0 \checkmark$

Else,  $\|f+g\|_p > 0$ .

Since  $x \rightarrow x^p$  convex,  $\left(\frac{f+g}{2}\right)^p \leq \frac{1}{2}(f^p + g^p)$  pointwise

$$(f+g)^p = f(f+g)^{p-1} + g(f+g)^{p-1}$$

$$\int f(f+g)^{p-1} \leq \|f\|_p \left( \int (f+g)^{(p-1)q} \right)^{\frac{1}{q}}$$

$$= \|f\|_p \left( \int (f+g)^p \right)^{\frac{1}{q}}$$

Similarly,

$$\int g(f+g)^{p-1} \leq \|g\|_p \left( \int (f+g)^p \right)^{\frac{1}{q}}$$

$$\therefore \int (f+g)^p \leq [\|f\|_p + \|g\|_p] \left( \int (f+g)^p \right)^{\frac{1}{q}}$$

$$\Rightarrow \left( \int (f+g)^p \right)^{1-\frac{1}{q}} \leq \|f\|_p + \|g\|_p \Rightarrow \boxed{\|f+g\|_p \leq \|f\|_p + \|g\|_p}$$