

Lecture 25 Synopsis  
 (Wed, 20 Apr)

The lecture covered a variety of topics.

First, regarding Lindelöf's  $\mu$ -function for  $I(s)$ . Cf.  
 Lec 24 p. (11) ff.

↑  
 note Lec 24, p. (13) lines 4-8

Thm

Consider  $f(s) \approx I(s)$  for  $\operatorname{Im}(s) \geq 1$ , say.

(a)  $\mu(\sigma) + (\sigma - \frac{1}{2}) = \mu(s - \sigma)$

(b)  $\mu(\sigma) = 0$ ,  $\sigma > 1$

(c)  $\mu(\sigma) \approx \frac{1}{2} - \sigma$ ,  $\sigma < 0$

(d)  $\mu(\sigma)$  is convex on every  $[a, b]$

(e)  $\mu(\sigma)$  is continuous on  $\operatorname{TR}$

(f)  $\mu(\sigma) \geq 0$

(g)  $\mu(\sigma)$  is monotonic decreasing

(h)  $\mu\left(\frac{1}{2}\right) \approx \frac{1}{4}$

(i) Lindelöf's conjecture is true  $\Leftrightarrow \mu\left(\frac{1}{2}\right) = 0$ .

P.F

(a) Lec 24 (15).

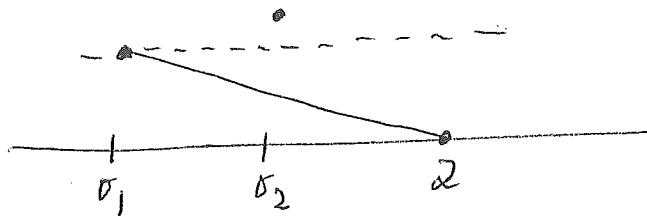
(b) Lec 24 (13) bot + (14).

(c) Combine (a)+(b). See Lec 24 (16).

(d) Lec 24 (12).

(2)

(e) Lec 24 (13) top:

(f) Know  $\mu(\sigma) = 0$ ,  $\sigma > 1$ . Hence  $\mu(\sigma) = 0$ ,  $\sigma \geq 1$ .Suppose  $\mu(\sigma_1) < 0$  with some  $\sigma_1 < 1$ . Take $\sigma_2 = 2$  and apply convexity over  $[\sigma_1, \sigma_2]$ .Get  $\mu\left(\frac{3}{2}\right) < 0$ . Contrad!(g) Know  $\mu(\sigma) \geq 0$ . And  $\mu(\sigma) = 0$ ,  $\sigma \geq 1$ .Suppose  $\sigma_1 < \sigma_2$  has  $0 \leq \mu(\sigma_1) < \mu(\sigma_2)$ .So,  $\sigma_2 < 1$ . Look at convexity over  $[\sigma_1, 2]$ .This violates convexity (at  $\sigma_2$ ).(h) Lec 24 p. (16) line 4, put  $\sigma = 1/2$ .

(i) Lec 24 p. (17).

Recall Lindelöf's Conjecture

$$\mu(\sigma) = \begin{cases} 0, & \frac{1}{2} \leq \sigma < \infty \\ \frac{1}{2} - \sigma, & -\infty < \sigma \leq \frac{1}{2} \end{cases}.$$

It is known that RH  $\Rightarrow$  Lindelöf Conjecture.

Lec 24  
16

## 2<sup>nd</sup> topic.

I briefly discussed the following thm.

### Thm

Let  $f(s) = \sum_{n=1}^{\infty} a_n \lambda_n^{-s}$  be a given generalized Dirichlet series with  $1 = \lambda_1 < \lambda_2 < \lambda_3 < \dots \rightarrow \infty$ . Suppose the series converges at  $s_0 \in \mathbb{C}$ .

Then:

- (a) the series conv uniformly on every Stolz angle

$$\left\{ |\operatorname{Arg}(s-s_0)| \leq \frac{\pi}{2} - \delta \right\};$$

- (b) the series conv uniformly on every "super" Stolz angle

$$\left\{ |t-t_0| \leq e^{M(\delta-\delta_0)} - 1 \right\}$$

$$(M > 0).$$

The proof (omitted here) is an interesting exercise. Of course, (a) is known already by Lec 21, p. ⑪ Fact 2. Concerning (b),

(4)

I simply remark: just study Lec 21, p. (17),  
line 7 when (wlog)  $\gamma_0 = 0$ . For  $\sigma > A$  big  
[but frozen], notice that:

$$\sigma \leq e^{M\sigma} \quad (M \geq 1 \text{ wlog})$$

$$|t| \leq e^{M\sigma} - 1 \leq e^{M\sigma}$$

$$|\omega| \leq 2e^{M\sigma} \text{ a priori}$$



get a  $\frac{\epsilon \cdot 2}{A} e^{-\sigma(\ln N - M)}$  term!

Needless to say, by a minor expungement  
and insertion (of a new " $\lambda$ "), we can  
actually allow ANY  $\lambda_i$  in the above Thm;  
we do not need  $\lambda_1 = 1$  ONLY  $\lambda_1 > 0$ .

Because of (3) Thm, Stolz angles or "super"  
Stolz angles are natural vehicles on which to  
discuss, e.g., identity theorems of the sort  
 $f_1(\xi_k) = f_2(\xi_k)$ , all  $k \geq 1 \Rightarrow a_{n1} = a_{n2}$ .

(5)

### 3<sup>rd</sup> topic.

We did a quick review of basic Fourier transforms and related analysis.

$$\hat{f}(p) \equiv \int_{-\infty}^{\infty} f(x) e^{-2\pi i p x} dx \quad p \in \mathbb{R}$$

$\int_{-\infty}^{\infty} |f(x)| dx < \infty$ , if piecewise  $C^1$  basically

$$\frac{f(x+0) + f(x-0)}{2} = \int_{-\infty}^{\infty} \hat{f}(p) e^{2\pi i p x} dp$$

$$RHS \equiv \lim_{R \rightarrow \infty} \int_{-R}^R \hat{f}(p) e^{2\pi i p x} dp$$


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$$\tilde{f}(u) \equiv \int_{-\infty}^{\infty} f(x) e^{-iux} dx$$

$$\frac{f(x+0) + f(x-0)}{2} = \frac{1}{2\pi} \int_{-\infty}^{\infty} \tilde{f}(u) e^{iux} du$$


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(6)

For "nice" functions  $f, g$  (real or complex) on  $\mathbb{R}$ , we define the convolution

$$H(x) = \int_{-\infty}^{\infty} f(t)g(x-t) dt$$

$H(x)$  is a reasonable function, due to

$$|H(x)| \leq \int_{-\infty}^{\infty} |f(t)| |g(x-t)| dt$$

$$\int_{-\infty}^{\infty} |H(x)| dx \leq \int_{-\infty}^{\infty} \left( \int_{-\infty}^{\infty} |f(t)| |g(x-t)| dt \right) dx$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |f(t)| |g(x-t)| dx dt$$

$\{$  Fubini's  $\}$

$$= \int_{-\infty}^{\infty} |f(t)| \left( \int_{-\infty}^{\infty} |g(x)| dx \right) dt$$

$$= \left( \int_{-\infty}^{\infty} |f(t)| dt \right) \left( \int_{-\infty}^{\infty} |g(x)| dx \right)$$

$\leftarrow \infty$

Often,  $f$  and  $g$  are initially kept in the Schwartz class  $S$ .

(7)

One easily checks that  $H(x)$  is continuous and bounded if either  $f$  or  $g$  is known to be bounded. This is true in all "sensible" cases.

↙ review (6) middle

What is also checked quite easily by Fubini and the key fact that

$$e^{i\theta} e^{i\phi} = e^{i(\theta+\phi)} \quad \begin{matrix} \theta \in \mathbb{R} \\ \phi \in \mathbb{R} \end{matrix}$$

is the relation from Fourier transform theory

$$\hat{H}(p) = \hat{f}(p) \hat{g}(p) \\ (\text{also } \tilde{H}(u) = \tilde{f}(u) \tilde{g}(u)) .$$

This is WHY the convolution  $H = f * g$  is so useful!

Another useful property goes as follows.  
Assume  $f, g, \hat{f}, \hat{g}$  are all "nice". Then,  
observe that:

(8)

$$\begin{aligned}
 & \int_{-\infty}^{\infty} f(x) \overline{g(x)} dx \\
 &= \int_{-\infty}^{\infty} f(x) \overline{\left[ \int_{-\infty}^{\infty} \hat{g}(p) e^{2\pi i p x} dp \right]} dx \\
 &= \int_{-\infty}^{\infty} f(x) \left[ \int_{-\infty}^{\infty} \overline{\hat{g}(p)} e^{-2\pi i p x} dp \right] dx \\
 &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x) \overline{\hat{g}(p)} e^{-2\pi i p x} dx dp \\
 &\quad (\text{by Fubini}) \\
 &= \int_{-\infty}^{\infty} \overline{\hat{g}(p)} \left[ \int_{-\infty}^{\infty} f(x) e^{-2\pi i p x} dx \right] dp \\
 &= \int_{-\infty}^{\infty} \hat{f}(p) \overline{\hat{g}(p)} dp .
 \end{aligned}$$

Thus,

$$\int_{-\infty}^{\infty} |f_1(x) - f_2(x)|^2 dx = \int_{-\infty}^{\infty} |\hat{f}_1(p) - \hat{f}_2(p)|^2 dp$$

for nice  $f_j$ . In particular:

$$\int_{-\infty}^{\infty} |f_j(x)|^2 dx = \int_{-\infty}^{\infty} |\hat{f}_j(p)|^2 dp .$$

This is the Plancherel formula.

(9)

Let  $\chi_E(x) = \begin{cases} 1, & x \in E \\ 0, & x \notin E \end{cases}$ . One easily checks:

$$\overbrace{\chi_{[-c, c]}}^{}(x) = 2 \frac{\sin cu}{u}$$

$$\overbrace{\max(0, b - |x|)}^{} = 2 \frac{1 - \cos bu}{u^2} = \frac{4 \sin^2(\frac{b}{2}u)}{u^2}$$

$$\{ u = 2\pi p \}$$

$$\overbrace{\chi_{[nc, nc]}}^{}(x) = \frac{\sin 2\pi pc}{\pi p}$$

$$\overbrace{\max(0, b - |x|)}^{} = \frac{1 - \cos 2\pi pb}{2\pi^2 p^2} = \frac{\sin^2(\pi pb)}{\pi^2 p^2}.$$

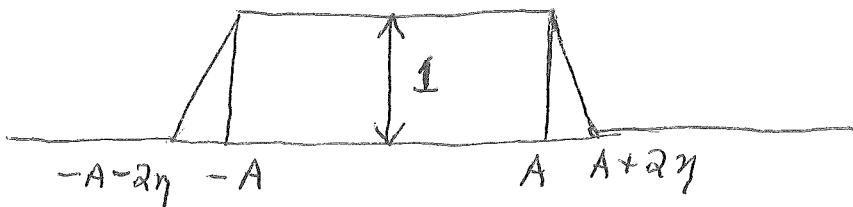
It will be convenient to consider the convolution

$$T(x) = \frac{1}{2\eta} \chi_{[-\eta, \eta]}(x) * \chi_{[-A-\eta, A+\eta]}(x).$$

THM

$A > 0, \eta > 0$ .  $T(x)$  as above. Then:

(1)  $T(x)$  is the trapezoid



(2)

$$\tilde{T}(x) = \frac{\cos(Au) - \cos((A+2\eta)u)}{\eta u^2}$$

(3)

$$\tilde{T}(x) = 2 \frac{\sin(\eta u) \sin((A+\eta)u)}{\eta u^2}$$

Pf

By ⑦ + ⑨,

$$\tilde{T}(x) = \frac{1}{2\eta} 2 \frac{\sin \eta u}{u} 2 \frac{\sin (A+\eta)u}{u}$$

$$= 2 \frac{\sin(\eta u) \sin((A+\eta)u)}{\eta u^2}, \quad (11)$$

so (3) is OK. Of course,

$$\cos(\theta - \phi) - \cos(\theta + \phi) = 2 \sin \theta \sin \phi$$



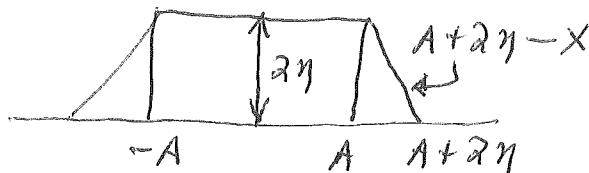
$$\begin{aligned} \cos(Au) - \cos((A+2\eta)u) &= 2 \sin((A+\eta)u) \sin(\eta u) \\ &= 2 \sin(\eta u) \sin((A+\eta)u), \end{aligned}$$

so (2) is OK too.

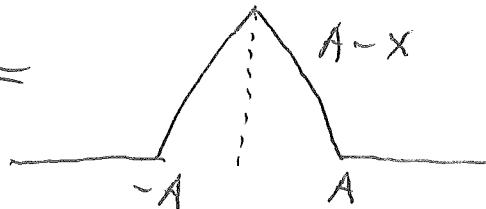
For (1), define  $g(x)$  to be the trapezoid you

(10). Look at:

$$2\eta g(x) \approx$$

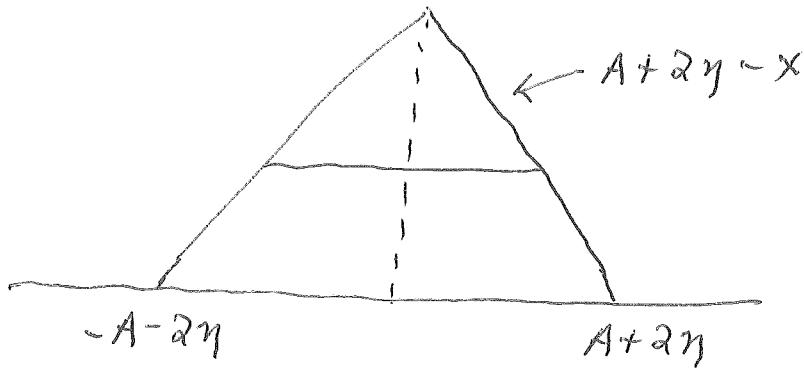


$$\max(0, A - |x|) =$$



(12)

$$2\eta g(x) + \max(0, A - |x|) =$$



$$= \max(0, A + 2\eta - |x|)$$



$$2\eta g(x) = \max(0, A + 2\eta - |x|) - \max(0, A - |x|)$$

$$\begin{aligned} 2\eta \tilde{g}(x) &= 2 \left[ \frac{1 - \cos((A+2\eta)u)}{u^2} \right] \\ &\sim 2 \left[ \frac{1 - \cos(Au)}{u^2} \right] \end{aligned}$$

$$= 2 \left[ \frac{\cos(Au) - \cos((A+2\eta)u)}{u^2} \right]$$

$$\Rightarrow \tilde{g}(x) = \frac{\cos(Au) - \cos((A+2\eta)u)}{2\eta u^2}.$$

By (2) on p. (10),

$$\tilde{f}(x) = \tilde{T}(x) \quad \text{(all } u \in \mathbb{R})$$

Apply (5) last line to this situation. Get

$$f(x) = T(x), \quad \text{each } x \in \mathbb{R}$$

since  $f$  and  $T$  are continuous on  $\mathbb{R}$ . Cf.  
 also the  $\tilde{f}$  counterpart of (8) line -4. In  
 any case, (1) is now true. ■

## 4<sup>th</sup> Topic

THM (an important estimate for  
Dirichlet polynomials) .

We have:

$$\int_{-\alpha}^{\alpha+H} \left| \sum_{j=1}^N b_j e^{-i\lambda_j t} \right|^2 dt = H \sum_{j=1}^N |b_j|^2 + \frac{O(1)}{\delta} \sum_{j=1}^N |b_j|^2$$

anytime

$$0 < \lambda_1 < \lambda_2 < \dots < \lambda_N$$

$$|\lambda_k - \lambda_j| \geq \underline{\delta}, \text{ all } k \neq j$$

$$b_j \in \mathbb{C}, \alpha \in \mathbb{R}, H > 0$$

The "implied" constant in  $O(1)$  is absolute;  
it can be taken to be  $\frac{4\pi}{\sqrt{3}}$ .

Pf

Some easy wlog's (giving same implied constant in  $O(1)$ ).

First one:  $\varphi = -\frac{H}{2}$ . Second one:  $H=2$ .

Must prove:

$$\int_{-1}^1 \left| \sum_{j=1}^N b_j e^{-i\lambda_j t} \right|^2 dt = 2 \sum_{j=1}^N |b_j|^2 + \frac{O(1)}{\delta} \sum_{j=1}^N |b_j|^2.$$

Take  $T(t)$  on ⑩ with  $A=I_J$ ,  $\eta=\eta$ . Let

$$J = \int_{-1}^1 \left| \sum b_j e^{-i\lambda_j t} \right|^2 dt.$$

Clearly

$$J \leq \int_{\mathbb{R}} \left| \sum b_j e^{-i\lambda_j t} \right|^2 T(t) dt$$

$$J \leq \sum_{j,k} b_j \overline{b_k} \int_{\mathbb{R}} T(t) e^{-i(\lambda_j - \lambda_k)t} dt$$

$$\{ \text{put } d_{jk} = \lambda_j - \lambda_k \}$$

(16)

$$\mathcal{F} \leq \sum_{j \neq k} b_j^* \bar{b}_k \frac{2}{\eta} \frac{\sin(\eta d_{jk}^*)}{d_{jk}^*} \frac{\sin[(1+\eta)d_{jk}^*]}{d_{jk}^*}$$

by (10)(3)

$$\begin{aligned} \mathcal{F} &\leq \sum_{j=1}^N |b_j^*|^2 \left\{ \frac{2}{\eta} \eta (1+\eta) \right\} \\ &\quad + \frac{2}{\eta} \sum_{j \neq k} b_j^* \bar{b}_k \frac{\sin(\eta d_{jk}^*)}{d_{jk}^*} \frac{\sin[(1+\eta)d_{jk}^*]}{d_{jk}^*} \end{aligned}$$

$$\begin{aligned} &\leq (2+2\eta) \sum_1^N |b_j^*|^2 \\ &\quad + \frac{2}{\eta} \sum_{j \neq k} \operatorname{Re}(b_j^* \bar{b}_k) \boxed{*} \boxed{*} \end{aligned}$$

$$\left\{ \text{but } 2\operatorname{Re}(b_j^* \bar{b}_k) \leq |b_j^*|^2 + |b_k|^2 \right\}$$

$$\leq (2+2\eta) \sum_1^N |b_j^*|^2 + \frac{1}{\eta} \sum_{j \neq k} \frac{|b_j^*|^2 + |b_k|^2}{d_{jk}^{*2}}$$

$$= (2+2\eta) \sum_1^N |b_j^*|^2 + \frac{2}{\eta} \sum_{j \neq k} \frac{|b_j^*|^2}{d_{jk}^{*2}}$$

(17)

$$\leq (2+2n) \sum_1^N |b_j|^2 + \frac{2}{\gamma} \sum_1^N |b_j|^2 \left( 2 \sum_{m=1}^{\infty} \frac{1}{m^2 \delta^2} \right)$$

↑

somewhat crudely  
 via  $|z_j - z_k| \geq 5 > 0$   
 for  $j \neq k$

$$= (2+2n) \sum_1^N |b_j|^2 + \frac{4}{\gamma \delta^2} \sum_1^N |b_j|^2 \frac{\pi^2}{6}$$

$$\left\{ \text{let } D = \frac{\pi}{\sqrt{3}} \right\}$$

$$= (2+2n) \sum_1^N |b_j|^2 + \frac{2D^2}{\gamma \delta^2} \sum_1^N |b_j|^2$$

$$= (2+2\gamma + \frac{2D^2}{\gamma \delta^2}) \sum_1^N |b_j|^2 .$$

To minimize RHS, take

$$\gamma = \frac{D}{\delta} .$$

(18)

Get:

$$\mathcal{F} \leq \left(2 + \frac{4D}{\delta}\right) \sum_j |b_j|^2$$

or

$$\int_{-1}^1 \left| \sum_{j=1}^N b_j e^{-i\lambda_j t} \right|^2 dt \leq 2 \sum_j |b_j|^2 + \frac{4\pi/\sqrt{3}}{\delta} \sum_j |b_j|^2.$$

This, of course, is the upper bound  
postulated on page (14).

(actually)

[The remainder of the proof was done in  
Lec 26, but we include it here!]

The lower bound for  $\mathcal{F}$  is similar but  
slightly harder. Must prove:

$$\mathcal{F} \geq \left(2 - \frac{4D}{\delta}\right) \sum_j |b_j|^2.$$

(19)

If  $\delta \leq 2D = \frac{2\pi}{\sqrt{3}}$ , matters are trivial.  
 So, wlog,  $\delta > 2D$ . Hence,  $\frac{1}{2} > \frac{D}{\delta}$ .

We consider  $T(t)$  on (10) with  $A = 1 - 2\gamma$ ,  $0 < \gamma < \frac{1}{2}$   
 and observe that  $A + 2\gamma = 1$ . Here

$$\begin{aligned}
 J &\geq \int_{-\infty}^{\infty} |T(t)| \left| \sum_j b_j e^{i\lambda_j t} \right|^2 dt \\
 &= \sum_j b_j \bar{b_k} \int_{-\infty}^{\infty} T(t) e^{-i(\lambda_j - \lambda_k)t} dt \\
 &= \sum_j b_j \bar{b_k} 2 \frac{\sin(\gamma d_{jk})}{\gamma d_{jk}} \frac{\sin((1-\gamma)d_{jk})}{d_{jk}} \\
 &= \sum_{j=1}^N |b_j|^2 (2 - 2\gamma) \\
 &\quad + \sum_{j \neq k} \operatorname{Re}(b_j \bar{b_k}) 2 \frac{\sin \gamma d_{jk}}{\gamma d_{jk}} \frac{\sin((1-\gamma)d_{jk})}{d_{jk}} \\
 &\quad \quad \quad \{ \text{as on (16)} \} \\
 &\geq (2 - 2\gamma) \sum_j |b_j|^2 - \sum_{j \neq k} (|b_j|^2 + |b_k|^2) \frac{1}{\gamma d_{jk}} \frac{1}{d_{jk}} \\
 &= (2 - 2\gamma) \sum_j |b_j|^2 - \frac{2}{\gamma} \sum_j |b_j|^2 \left( \sum_{k \neq j} \frac{1}{d_{jk}^2} \right) \\
 &\quad \quad \quad \{ \text{as on (16) bottom} \}
 \end{aligned}$$

$$\geq (\alpha - 2\gamma) \sum_1^N |b_j|^2 = \frac{\alpha}{\gamma} \sum_1^N |b_j|^2 \frac{1}{\delta^2} \alpha \left(\frac{\pi^2}{6}\right)$$

(20)

$$\left\{ \text{cf. (17) middle} \right\} \left\{ D = \frac{\pi}{\sqrt{3}} \right\}$$

$$= \sum_1^N |b_j|^2 \left( \alpha - 2\gamma - \frac{\alpha D^2}{\gamma \delta^2} \right).$$

Take  $\gamma = \frac{D}{\delta}$ . Since  $\frac{1}{2} > \frac{D}{\delta}$ ,  $\gamma$  is admissible.

Get

$$J \geq \alpha \sum_1^N |b_j|^2 - \frac{4D}{\delta} \sum_1^N |b_j|^2,$$

with  $4D = \frac{4\pi}{\sqrt{3}}$ , parallel to (18) lines 4-5. This is the lower bound promised.  $\blacksquare$

Let  $C$  be the constant in the  $O(1)$  on (14). The theorem on (14) is very closely related to the generalized Hilbert inequality.

$$(*) \quad \left| \sum_{j \neq k} \frac{z_j \bar{z}_k}{\lambda_j - \lambda_k} \right| \leq \frac{C/2}{\delta} \sum_{j=1}^N |\lambda_j|^2 \quad (z_j \in \mathbb{C}).$$

One readily checks that  $(*) \Rightarrow$  thm on (14). Selberg has noted [in a very slick proof] that the thm on (14)  $\Rightarrow$   $(*)$ .

By choosing majorant/minorant functions more sophisticated than trapezoids, one finds that the best  $C$  is  $2\pi$ .

Note:

$$\frac{4\pi}{\sqrt{3}} = 3\pi(1.1547^+),$$

which is not too bad!

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