

Lecture 27 Synopsis
 (Wed, 27 Apr)

We seek to develop the famous thm of Bohr-Landau about zeros of $\zeta(s)$ to the right of $\operatorname{Re}(s) = \frac{1}{2}$.

Recall, from Lec 24, (19)(20), that we had

$N(u; T_1, T_2) = \# \text{ of zeros of } f(s) \text{ on } R \cup \partial R$
 having abscissa} > u \text{ (and counted
 WITH multiplicity)}

wherein $R = (\alpha, \beta) \times (T_1, T_2)$; $f(s)$ is analytic on $R \cup \partial R$, $f(\beta + it) \neq 0$, $f(\alpha + iT_1) \neq 0$, $f(\alpha + iT_2) \neq 0$.
 One defines $\phi(s) = \operatorname{Log} f(s)$ via horizontal analytic continuation starting at $\sigma = \beta$ insofar as $t \neq$ ordinate of a zero of f ; otherwise by an obvious right continuity from above.

In this framework, we got Littlewood's formula

$$-\frac{1}{2\pi i} \oint_{\partial R} \phi(s) ds = \int_{\alpha}^{\beta} N(u; T_1, T_2) du$$

and, then, the simplified version in Lec 24, (25).

(2)

The Bohr-Landau theorem will arise by specializing $f(s)$ to be $I(s)$ in the foregoing — and playing with appropriate α and β .

Not-too-surprisingly, matters will need to be looked at along the way with the aid of some of our previously obtained estimates.

1914

THEOREM (basic form of Bohr-Landau thm)

Consider $I(s)$. For $T \geq 2$, say, [not the ordinate of a zero of I], we have:

$$\boxed{N\left(\frac{1}{2} + \varepsilon; h, T\right) = O_\varepsilon(T)}$$

for each $\varepsilon > 0$. Here h is a tiny number such that $I \neq 0$ on $\{\operatorname{Re}(s) \geq 0, 0 \leq \operatorname{Im}(s) \leq h\}$; see Lec 11, p. 27 and Lec 13, pp. 6 (top) + 9 (top).

We stress here that, in toto, we have

$$N(0; h, T) \sim \frac{T}{2\pi} \ln\left(\frac{T}{2\pi e}\right)$$

by Lec 15, pp. 29 + 22 (box). As such, by the functional equation of I , only 0% of the complex zeros of I can lie outside $|\operatorname{Re}(s) - \frac{1}{2}| \leq \varepsilon$.

(3)

The relevant "PREVIOUSLY OBTAINED" estimates referenced on $\sqrt{2}$ are those found on

page ⑧ of Lec 15 (a priori upper bound)

page ⑬ of Lec 15 (the partial fraction thing)

page ⑭ of Lec 25 (L_2 estimate)

page ⑪ of Lec 26 (Hardy-Littlewood estimate).

In connection with the first two — from Lec 15 — we note the following :

FACT

Keep $T \geq 2$, say, and not the ordinate of a zero of I . Use the standard UP AND ACROSS definition of $\log I(r)$ beginning at some point $A \in \mathbb{R}$, $A \gg 1$. Then:

$$\arg I(\sigma + iT) = O(\ln T),$$

for all $-1 \leq \sigma \leq 2$.

PF

by the Dirichlet series!

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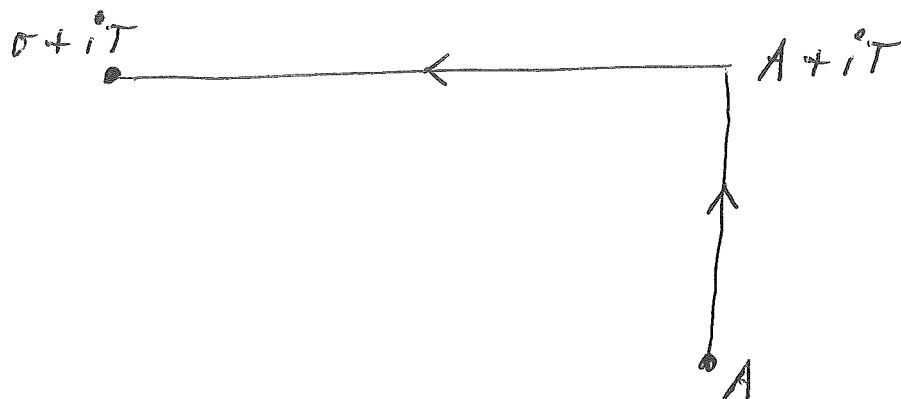
We know $|\log \zeta(s)| = D_\eta(1)$, $\sigma \geq 1 + \eta$.

Use

[hence also for
 $\arg \zeta(s)$]

$$N[|y-t| \leq 1] = O(\ln t)$$

$$\frac{\zeta'(s)}{\zeta(s)} = O(\ln t) + \sum_{\substack{p \\ |y-t| \leq 1}} \frac{1}{s-p} \quad \left\{ -1 \leq \sigma \leq 2 \right\}$$



$$\arg \zeta(\sigma+iT) = \operatorname{Im} \int_A^\sigma \frac{\zeta'}{\zeta}(u+iT) du$$

$$+ \operatorname{Im} \log \zeta(A+iT)$$

this is $O(\tau^{-A})$
by the Dirichlet series for $\log \zeta$

{we can let $A \rightarrow \infty$ }

(5)

$$\begin{aligned}\operatorname{Arg} f(\sigma + iT) &= -\operatorname{Im} \int_{\sigma}^{\infty} \frac{T}{z}(u+iT) du \\ &= -\operatorname{Im} \int_{\sigma}^{\infty} -\operatorname{Im} \int_2^{\infty} \frac{T}{f}(u+iT) du\end{aligned}$$

$O(\alpha^{-u})$ by the
Dirichlet series
for $-\frac{f'}{f}$

$$\begin{aligned}&= -\operatorname{Im} \int_{\sigma}^{\infty} \left\{ O(\ln T) + \sum_p \frac{1}{u+iT-p} \right\} du \\ &\quad |y-T| \leq 1 \\ &\quad + O(1)\end{aligned}$$

$$= O(\ln T)$$

$$-\sum_p \left\{ \operatorname{Arg}(z+iT-p) - \operatorname{Arg}(\sigma+iT-p) \right\}$$

$$|y-T| \leq 1$$

$\operatorname{Arg} = \text{ordinary principal value}$

$$= O(\ln T) + O(\ln T) + O(1) \quad (6)$$

$$= O(\ln T)$$

{ much like Lec 15, p. 28 } . \blacksquare

By Littlewood's identity, Lec 24, p. 25, know:

$$\frac{1}{2} \leq \gamma < \beta \leq 2$$



$$2\pi \int_{\alpha}^{\beta} N(\sigma; T_1, T_2) d\sigma$$

$$= \int_{T_1}^{T_2} \ln |\zeta(\gamma+it)| dt$$

$$- \int_{T_1}^{T_2} \ln |\zeta(\beta+it)| dt$$

$$+ O[(\beta-\gamma) \ln T_2] \quad \leftarrow \begin{matrix} \text{by FACT} \\ \text{on ③} \end{matrix}$$

at least for T_j which are not the ordinates of ζ -zeros. We'll keep $T_2 > T_1 \geq 2$. Cf. also h on page 2.

(7)

Must focus on

$$\int_{T_1}^{T_2} \ln |I(\sigma + it)| dt$$

(since β will be taken $\geq \frac{3}{2}$ later) •

see (15) middle

We propose to look first at

$$\int_{T/2}^T \underline{|I(\sigma + it)|^2} dt$$

TRICK

with $T \geq 1000$ and $\frac{1}{2} \leq \sigma \leq 2$ •

Use H-L estimate from Lec 26 (11), $\sigma_0 = \frac{1}{10}$ •

Take $C = \pi$ • We get:

$$I(\sigma + it) = \sum_{n \leq T} n^{-\sigma - it} - \frac{T^{1-\sigma - it}}{1-\sigma - it} + O(T^{-5})$$

for $t \in [\frac{1}{2}T, T]$ •

(8)

So,

$$f(\sigma + it) = \sum_{n \leq T} n^{-\sigma} e^{-it \ln n} + \frac{O(T^{1-\sigma})}{T} + O(T^{-\sigma})$$

$$= \sum_{n \leq T} n^{-\sigma} e^{-it \ln n} + O(T^{-\sigma})$$

for $\frac{1}{2}\tau \leq t \leq T$. We'll write this as

$$f(\sigma + it) = \underbrace{\sum}_{\equiv} + R.$$

Hence,

$$|f(\sigma + it)|^2 = (\sum + R)(\bar{\sum} + \bar{R})$$

$$= \sum \bar{\sum} + 2 \operatorname{Re}(R \bar{\sum}) + |R|^2$$

$$= \sum \bar{\sum} + O(1) T^{-\sigma} |\sum| + O(1) T^{-2\sigma}$$

for $t \in [\frac{1}{2}\tau, T]$ and $\frac{1}{2} \leq \sigma \leq 2$.

(9)

Put:

$$\sum = \sum_{n \leq T} \tilde{n}^{-\sigma} e^{-i\lambda_n t} \quad \text{with}$$

$$\lambda_n \equiv \ln n \rightarrow$$

and then use Lec 25 p. ⑭ (the L_2 estimate).
 Here, of course, $n > m \Rightarrow$

$$\begin{aligned} \lambda_n - \lambda_m &= \ln n - \ln m \\ &= \frac{1}{\tilde{n}}(n-m), \quad \tilde{n} \in (m, n) \\ &\approx \frac{1}{T} \quad . \end{aligned}$$

We can apply Lec 25 ⑭ with $\delta = \frac{1}{T}$. So,

$$\begin{aligned} \int_{T/2}^T |\sum|^2 dt &= \frac{T}{2} \sum_{n \leq T} \tilde{n}^{-2\sigma} \\ &\quad + \frac{O(1)}{T} \sum_{n \leq T} \tilde{n}^{-2\sigma} \end{aligned}$$

{this may be improvable, but we
prefer to stay with a crude bound!}

(10)



$$\int_{T/2}^T |\sum|^{\sigma} dt = O(T) \sum_{n \leq T} n^{-\sigma} .$$

Suppose now that $\sigma > \frac{1}{2}$. In that case, we go further and get

$$\begin{aligned} \int_{T/2}^T |\sum|^{\sigma} dt &= O(T) I(2\sigma) \\ &= O(T) \frac{1}{2\sigma - 1} . \end{aligned}$$

$$\boxed{\sigma \in (\frac{1}{2}, 2]}$$

At the same time, (see ⑧ bottom)

$$\begin{aligned} \int_{T/2}^T T^{-\sigma} |\sum| dt &\leq \left\{ \int_{T/2}^T T^{-2\sigma} dt \right\}^{\frac{1}{2}} \\ &\quad \cdot \left\{ \int_{T/2}^T |\sum|^{\sigma} dt \right\}^{\frac{1}{2}} \end{aligned}$$

{by Cauchy-Schwarz}

$$\leq \sqrt{T^{1-2\sigma}} \sqrt{\frac{O(T)}{2\sigma - 1}}$$

$$= O(1) \frac{T^{1-\sigma}}{\sqrt{2\sigma - 1}} = O(1) \frac{T^{1-\sigma}}{2\sigma - 1} .$$

(11)

Referring to ⑧ bottom again, we get

$$\begin{aligned} \int_{T/2}^T |f(\sigma + it)|^2 dt \\ = \frac{o(T)}{\sigma - 1} + \frac{o(T^{1-\sigma})}{\sigma - 1} + o(T^{1-2\sigma}) \end{aligned}$$

for $\frac{1}{2} < \sigma \leq 2$. Accordingly (via $2\sigma \geq 1$):

$$\int_{T/2}^T |f(\sigma + it)|^2 dt \leq \frac{o(1)}{\sigma - \frac{1}{2}} T$$

for $\frac{1}{2} < \sigma \leq 2$. The case $\sigma = \frac{1}{2}$ is also OK, but [obviously] not very informative.

Suppose NEXT that we only know $\sigma \geq \frac{1}{2}$.

Observe that ⑩ line 1 is still usable.

Being totally crude, we can say:

$$\sum_{n \leq T} n^{-2\sigma} \leq \sum_{n \leq T} \frac{1}{n} \leq C + \ln T.$$

(12)

Hence,

$$\int_{T/2}^T |\sum|^2 dt = O(T \ln T)$$

{ can then mimic (10) bot + (11) top }



$$\begin{aligned} \int_{T/2}^T |J(\sigma + it)|^2 dt &= O(T \ln T) \\ &\quad + O(1) T^{1-\sigma} \sqrt{\ln T} \\ &\quad + O(T^{1-2\sigma}) \end{aligned}$$

$$\int_{T/2}^T |J(\sigma + it)|^2 dt \leq O(1) T \ln T$$

for $\frac{1}{2} \leq \sigma \leq 2$.

We have obtained the above box, and
 (11) box, assuming $T \geq 1000$ (see (7)).

(13)

THEOREM (standard a priori estimate)

For $T \geq 3$ and $\sigma \in [\frac{1}{2}, 2]$, we have:

$$\int_{\frac{1}{2}}^T |\mathcal{J}(\sigma+it)|^2 dt = O(T) \min \left\{ \ln T, \frac{1}{\sigma - \frac{1}{2}} \right\}.$$

PF

Matters are obvious for $3 \leq T \leq 10^6$. In fact, here,

$$\begin{aligned} \min \left\{ \ln T, \frac{1}{\sigma - \frac{1}{2}} \right\} &\stackrel{\forall}{=} \min \left\{ \ln 3, \frac{1}{3/2} \right\} \\ &= \frac{2}{3} \end{aligned}$$

and it suffices to adjust the implied constant in $O(T)$.

For $T > 10^6$, choose ℓ so that

$$\frac{T}{2^{\ell+1}} \leq 1000 < \frac{T}{2^\ell}.$$

Apply (11) box and (12) box to $\frac{T}{2^k}$ for $k \in [0, \ell]$.
Add! Get:

$$\int_{1000}^T |J(\sigma + it)|^2 dt = O(T) \min \left\{ \ln T, \frac{1}{\sigma - \frac{1}{2}} \right\}.$$

To replace 1000 by 2, repeat the observation used for $3 \leq T \leq 10^6$. \blacksquare

With more work, one can prove that:

$$\int_2^T |J(\sigma + it)|^2 dt \sim T J(2\sigma), \quad \sigma > \frac{1}{2}$$

and

$$\int_2^T |J(\frac{1}{2} + it)|^2 dt \sim T \ln T.$$

See Titchmarsh, Theory of $J(s)$, first few sections in chapter 7. We won't need these more precise results.

— — —

On p. ⑥ bottom, fix $T_1 \in [2, 2.5]$ to be well away from the ordinate of any I -zero.

Since numerical work shows that

$$\rho_1 = \frac{1}{2} + i[14.134725^+],$$

Lec 13 ⑤
Lec 22 ①

one can declare $T_1 = 2$. We prefer not to use this, however.

Keep $T = T_2 \geq 3$ and then take

$$\frac{1}{2} < \alpha \leq 1 \quad \text{and} \quad \underline{\beta = 2} \text{ (say).}$$

One gets:

$$\begin{aligned} 2\pi \int_{\alpha}^2 N(\alpha; T_1, T) d\alpha &= \int_{T_1}^T \ln |I(\alpha + it)| dt \\ &\quad - \int_{T_1}^T \ln |I(2 + it)| dt \\ &\quad + O(\ln T). \end{aligned}$$

BABY LEMMA

(a) for $x \in \mathbb{R}$, $e^x \geq 1+x$

(b) for $y \geq 0$, $\ln y \leq \frac{1}{2}(y^2 - 1)$.

\uparrow clearly very crude

Pf

We give 2 proofs of (a). The first notes that $g(x) = e^x$ is concave upward since $g'' > 0$. Hence $g(x)$ sits on or above the tangent line at each point x_0 . Take $x_0 = 0$. Get: $g(x) \geq 1+x$ by inspection.

The 2nd proof is more boring. For $x > 0$, apply mean value thm to get

$$e^x - 1 = e^{\tilde{x}}(x-0) = e^{\tilde{x}} \cdot x, \quad 0 < \tilde{x} < x$$

$$e^x - 1 \geq e^0 \cdot x = x. \quad \text{OK}$$

For $x < 0$, apply mean value theorem to get

$$1 - e^x = e^{\tilde{x}}(0-x) = (-x)e^{\tilde{x}}, \quad x < \tilde{x} < 0$$

$$1 - e^x \leq (-x)e^0$$

$$e^x - 1 \geq x. \quad \text{OK}$$

In (b), wlog $y > 0$. Put $y = e^u$ with $u \in \mathbb{R}$.

Must check that

(17)

$$u \leq \frac{1}{2}(e^{2u} - 1)$$

$$2u \leq e^{2u} - 1$$

$$e^{2u} \geq 1 + 2u$$

But this is obvious by (a). ■

By BABY LEMMA, then,

$$\begin{aligned} \int_{T_1}^T \ln|J(\sigma+it)| dt &\leq \frac{1}{2} \int_{T_1}^T (|J(\sigma+it)|^2 - 1) dt \\ &\leq \frac{1}{2} \int_{T_1}^T |J(\sigma+it)|^2 dt \\ &\leq \frac{1}{2} \int_2^T |J(\sigma+it)|^2 dt \\ &\leq \frac{O(T)}{\sigma - \frac{1}{2}} \quad \text{by (13)}. \end{aligned}$$

(18)

In addition,

$$\log \zeta(\alpha + it) = \sum_{n=2}^{\infty} \frac{1(n)}{\ln n} n^{-\alpha - it}$$

$$\begin{aligned} \int_{T_1}^T \ln |\zeta(\alpha + it)| dt &= \operatorname{Re} \int_{T_1}^T \log \zeta(\alpha + it) dt \\ &= \operatorname{Re} \int_{T_1}^T \sum_{n=2}^{\infty} \frac{1(n)}{\ln n} n^{-\alpha - it} dt \\ &= O(1) \sum_{n=2}^{\infty} \frac{1(n)}{\ln n} n^{-\alpha} \left[\frac{e^{-it \ln n}}{-i \ln n} \right]_{T_1}^T \\ &= O(1) \sum_{n=2}^{\infty} \frac{1(n)}{\ln n} n^{-\alpha} \frac{1}{\ln n} \\ &= O(1) \cdot \boxed{\quad} \end{aligned}$$

Page (15) bottom then gives:

$$2\pi \int_{-\alpha}^{\alpha} N(\sigma; T_1, T) d\sigma \leq \frac{O(T)}{\alpha - \frac{1}{2}} + O(1) + O(\ln T)$$

$$\leq \frac{O(T)}{\alpha - \frac{1}{2}} \boxed{\quad} .$$

↑ here $\alpha \in (\frac{1}{2}, 1]$

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Write $\alpha = \frac{1}{2} + 2\omega$, $0 < \omega \leq \frac{1}{4}$. Notice that

$$N\left(\frac{1}{2} + 2\omega; T_1, T\right) \leq \frac{1}{\omega} \int_{\frac{1}{2} + \omega}^{\frac{1}{2} + 2\omega} N(\sigma; T_1, T) d\sigma$$

since $N(\sigma; T_1, T)$ is monotonic decreasing in σ .

Thus:

$$\begin{aligned} N\left(\frac{1}{2} + 2\omega; T_1, T\right) &\leq \frac{1}{\omega} \int_{\frac{1}{2} + \omega}^{\frac{1}{2}} N(\sigma; T_1, T) d\sigma \\ &\leq \frac{1}{\omega} \frac{\Omega(T)}{\omega} = \frac{4}{(2\omega)^2} \Omega(T). \end{aligned}$$

In other words,

$$N(\alpha; T_1, T) \leq \frac{\Omega(T)}{\left(\alpha - \frac{1}{2}\right)^2}, \quad \frac{1}{2} < \alpha \leq 1.$$

This proves the THM on p. ②. ■

Because our use of (16)(b) was so crude, one suspects that the foregoing box can be improved in the α -aspect.

 cf. also (18) bottom

(20)

This is indeed correct. We claim, in fact, that

$$\int_{T_1}^T \ln |\Im(\alpha + it)| dt = O(T) \ln \left(\frac{1}{\alpha - \frac{1}{2}} \right).$$

On ⑯, this leads to

$$N(\alpha; T_1, T) \leq O(T) \frac{1}{\alpha - \frac{1}{2}} \ln \left(\frac{1}{\alpha - \frac{1}{2}} \right).$$

There are 2 approaches to line ⑰.
proving

Method I

By ⑮(bottom), wlog $\alpha \in (\frac{1}{2}, \frac{3}{5}]$.

Under this hypothesis, we have $r \equiv \frac{1}{\alpha - \frac{1}{2}} \geq 10$.

Notice that $\ln y \leq \frac{1}{2\lambda} (y^{2\lambda} - 1)$ for $0 < \lambda \leq 1$
[and $y \geq 0$]. See ⑯. We'll keep $\lambda < 1$. Get:

(21)

$$\begin{aligned}
 \int_{T_1}^T \ln|z| dt &\leq \frac{1}{2\lambda} \int_{T_1}^T (|z|^{2\lambda} - 1) dt \\
 &\leq \frac{1}{2\lambda} \int_{T_1}^T |z|^{2\lambda} dt \\
 &\leq \frac{1}{2\lambda} \left(\int_{T_1}^T |z|^{2\lambda \frac{1}{1-\lambda}} dt \right)^{1-\lambda} \left(\int_{T_1}^T 1 dt \right)^{\lambda-1}
 \end{aligned}$$

{ by Hölder's inequality }

$$\begin{array}{l}
 p = \frac{1}{\lambda} \\
 q = \frac{1}{1-\lambda}
 \end{array}$$

$$\begin{aligned}
 &\leq \frac{1}{2\lambda} \left(\frac{O(T)}{q - \frac{1}{2}} \right)^{1-\lambda} T^{\lambda-1} \quad \text{by (13)} \\
 &\leq \frac{C}{\lambda} \left(\frac{1}{q - \frac{1}{2}} \right)^{\lambda} T \\
 &= \frac{C}{\lambda} r^\lambda \cdot T
 \end{aligned}$$

Put $\lambda = \frac{1}{\ln r}$. This is admissible since $r \geq 10$.

Get:

$$\begin{aligned}
 \int_{T_1}^T \ln|z(q+it)| dt &\leq C(\ln r) e^T \\
 &\leq O(T) \ln \left(\frac{1}{q - \frac{1}{2}} \right) \quad . \quad \text{OK}
 \end{aligned}$$

Method II

We use Jensen's inequality common in probability and measure theory.

To recall it, let $\Phi(v)$ be non-negative and convex on \mathbb{R} . (Hence Φ is automatically continuous.) Let Y be an extended real-valued random variable on a probability space (\mathcal{X}, μ) . Assume that $\mathbb{E}(Y)$ exists (i.e. that $Y \in L_1(\mu)$). Then:

$$\boxed{\Phi(\mathbb{E}(Y)) \leq \mathbb{E}(\Phi(Y))} \quad \boxed{\boxed{}}$$

One simply approximates $\mathbb{E}(Y)$ by an obvious Riemann sum and uses

$$\left\{ \begin{array}{l} \Phi\left(\sum_{j=1}^N t_j v_j\right) \leq \sum_{j=1}^N t_j \Phi(v_j) \\ \text{for } t_j \in [0, 1] \text{ with } t_1 + \dots + t_N = 1 \end{array} \right\} .$$

Put $\Phi(v) = \exp(v)$ to get

$$\exp\left(\frac{1}{H} \int_0^H f(t) dt\right) \leq \frac{1}{H} \int_0^H \exp[f(t)] dt$$

anytime $f \in L_1[0, H]$.

(23)

A trivial specialization gives :

$$\begin{aligned}
 & \exp \left(\frac{1}{T-T_1} \int_{T_1}^T 2 \ln |I(r+it)| dt \right) \\
 & \leq \frac{1}{T-T_1} \int_{T_1}^T |I(r+it)|^2 dt \\
 & \leq \frac{1}{T-T_1} \frac{O(T)}{r-\frac{1}{2}} \quad \text{by (13)} \\
 & \leq \frac{O(1)}{r-\frac{1}{2}}
 \end{aligned}$$

Accordingly :

$$\begin{aligned}
 & \frac{1}{T-T_1} \int_{T_1}^T 2 \ln |I(r+it)| dt \\
 & \leq \ln \left(\frac{\beta}{r-\frac{1}{2}} \right) \quad \left\{ \begin{array}{l} \text{some} \\ \beta \geq 1 \end{array} \right\}
 \end{aligned}$$



$$\int_{T_1}^T \ln |I(r+it)| dt \leq O(T) \ln \left(\frac{1}{r-\frac{1}{2}} \right)$$

for $r \in (\frac{1}{2}, 1]$. OK

(24)

Incidentally, observe how (13) THM gets applied in both methods I + II. Switching to $\ln T$ in place of $\frac{1}{q-\frac{1}{2}}$ produces the following:

method I

get $\frac{C}{\lambda} (\ln T)^{\lambda} T$ on (21) middle

$$\Rightarrow \text{take } \lambda = \frac{1}{\ln \ln T} \quad (T \text{ giant})$$

$$\Rightarrow \int_{T_1}^T \ln |\mathcal{I}(q+it)| dt = O(T \ln \ln T)$$

uniformly for $q \in [\frac{1}{2}, 1]$

method II

get $\leq O(1) \ln T$ on (23) middle

$$\Rightarrow \frac{1}{T-T_1} \int_{T_1}^T 2 \ln |\mathcal{I}| dt \leq \ln(B \ln T)$$

$$\Rightarrow \int_{T_1}^T \ln |\mathcal{I}(q+it)| dt = O(T \ln \ln T)$$

uniformly for $q \in [\frac{1}{2}, 1]$

HENCE:

$$\boxed{\int_{T_1}^T \ln |\mathcal{I}(q+it)| dt = O(T) \ln \left[\min \left(\ln T, \frac{1}{q-\frac{1}{2}} \right) \right].}$$

Closing Remark.

The estimate

$$N\left(\frac{1}{2} + \varepsilon; T_1, T\right) = O(T) \frac{1}{\varepsilon} \ln\left(\frac{1}{\varepsilon}\right)$$

mentioned on p. (20) (line 4) was obtained by Littlewood in 1924. Proc. Cambro. Philos. Soc. 22 (1924)

It is possible to expunge the term $\ln \frac{1}{\varepsilon}$.

This was shown by A. Selberg around 1942 with the aid of some fundamentally new ideas. By use of a so-called mollifier method, Selberg was able to demonstrate that

$$\int_{-\frac{1}{2}}^{\frac{1}{2}} N(\sigma; T_1, T) d\sigma = O(T) .$$

Compare (15) (bottom) + (18) (middle).

See also Titchmarsh, Theory of $I(s)$, around § 9.24.