

(July 4, 2015)

## From moments to pointwise estimates

Paul Garrett garrett@math.umn.edu http://www.math.umn.edu/~garrett/

[0.0.1] **Theorem:** For a degree  $n$   $L$ -function  $L(s)$  with suitable analytic continuation and functional equation, a second-moment bound

$$\int_0^T |L(\tfrac{1}{2} + it)|^2 dt \ll T^A \quad (\text{for all } T \gg 1)$$

implies a pointwise bound

$$L(\sigma_o + it) \ll_{\sigma_o, \varepsilon} (1 + |t|)^{\frac{A}{2} + \varepsilon} \quad (\text{for all } \sigma_o \geq \tfrac{1}{2} \text{ and for all } \varepsilon > 0)$$

[0.0.2] **Remark:** The convexity bound has exponent  $\frac{n}{4}$ . Thus, a moment bound with  $A < \frac{n}{2}$  would produce a subconvex bound.

*Proof:* Fix  $\sigma_o > \frac{1}{2}$ . For  $0 < t_o \in \mathbb{R}$ , let  $s_o = \sigma_o + it_o$ . Let  $R$  be the rectangle in  $\mathbb{C}$  with vertices  $\frac{1}{2} \pm iT$  and  $2 \pm iT$ , for  $T > t_o$ . By Cauchy's theorem,

$$L(s_o)^2 = \frac{1}{2\pi i} \int_R \frac{e^{(s-s_o)^2}}{s-s_o} \cdot L(s)^2 ds$$

Since the  $L$ -function has polynomial vertical growth, we can push the top and bottom of  $R$  to infinity, giving

$$\begin{aligned} L(s_o)^2 &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{e^{((\frac{1}{2}-\sigma_o)+i(t-t_o))^2}}{(\frac{1}{2}-\sigma_o)+i(t-t_o)} \cdot L(\tfrac{1}{2} + it)^2 dt + \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{e^{((2-\sigma_o)+i(t-t_o))^2}}{(2-\sigma_o)+i(t-t_o)} \cdot L(2 + it)^2 dt \\ &= \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{e^{((\frac{1}{2}-\sigma_o)+i(t-t_o))^2}}{(\frac{1}{2}-\sigma_o)+i(t-t_o)} \cdot L(\tfrac{1}{2} + it)^2 dt + O(1) \quad (\text{for } \tfrac{1}{2} < \sigma_o < 1) \end{aligned}$$

The part of the integral where  $|t - t_o| \geq t_o$  is visibly  $\ll_{n, \sigma_o} e^{-t_o}$ :

$$|e^{((\frac{1}{2}-\sigma_o)+i(t-t_o))^2}| = e^{(\frac{1}{2}-\sigma_o)^2 - (t-t_o)^2} \ll_{\sigma_o} e^{-t_o^2/2} e^{-(t-t_o)^2/2} \ll e^{-t_o} \quad (\text{for } |t - t_o| \geq t_o)$$

Squaring the convexity bound for  $L(\frac{1}{2} + it)$  gives

$$L(\tfrac{1}{2} + it)^2 \ll |t|^{\frac{n}{2} + \varepsilon} \quad (\text{for all } \varepsilon > 0)$$

Thus,

$$\int_{2t_o}^{\infty} \frac{e^{(\frac{1}{2}-\sigma_o)+i(t-t_o))^2}}{(\frac{1}{2}-\sigma_o)+i(t-t_o)} \cdot L(\tfrac{1}{2} + it)^2 dt \ll_{\sigma_o} e^{-t_o^2/2} \int_{2t_o}^{\infty} e^{-(t-t_o)^2/2} t^{\frac{n}{2} + \varepsilon} dt \ll_{n, \varepsilon} e^{-t_o}$$

The other half of the tail, where  $t < 0$ , is estimated similarly, and more easily. For  $0 < t < 2t_o$ , use the moment estimate, and the trivial estimate

$$\frac{e^{((\frac{1}{2}-\sigma_o)+i(t-t_o))^2}}{(\frac{1}{2}-\sigma_o)+i(t-t_o)} \ll_{\sigma_o} e^{(\frac{1}{2}-\sigma_o)^2 - (t-t_o)^2} \ll_{\sigma_o} 1$$

Then

$$\int_0^{2t_o} \frac{e^{(\frac{1}{2}-\sigma_o)+i(t-t_o))^2}}{(\frac{1}{2}-\sigma_o)+i(t-t_o)} \cdot L(\tfrac{1}{2} + it)^2 dt \ll_{\sigma_o} \int_0^{2t_o} |L(\tfrac{1}{2} + it)|^2 dt \ll t_o^A$$

Thus,

$$L(s_o)^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{e^{(\frac{1}{2}+it-s_o)^2}}{\frac{1}{2}+it-s_o} \cdot L(\tfrac{1}{2} + it, f)^2 dt + O(1) \ll_{n, \sigma_o} t_o^A$$

Then a standard convexity argument gives the asserted  $|t_o|^{\frac{A}{2} + \varepsilon}$  on  $\sigma_o = \frac{1}{2}$ , for every  $\varepsilon > 0$ . ///