

# Hilbert-Type Lemma with Möbius Coefficients, Numerical Calibration, and Extended NB/BD Criterion Towards the Riemann Hypothesis

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## Abstract

We establish a weighted Hilbert-type lemma for Möbius-weighted coefficients, proving that off-diagonal contributions in the associated normal equations are suppressed by a logarithmic factor. As a consequence, the Nyman–Beurling/Báez-Duarte (NB/BD) criterion remains stable, and the distance  $d_N$  tends to zero. Using a disjoint train/test grid with a zeta-weighted target, numerical experiments up to  $N = 20,000$  show a clear decay of mean square error (MSE). A regression of the form  $\log(\text{MSE}) = \alpha - \theta \log \log N$  on  $N \in \{8k, 12k, 16k, 20k\}$  yields  $\hat{\theta} \approx 7.21$  with a 95% CI  $[5.77, 8.65]$  on our dataset, consistent with the theoretical expectation  $\theta > 0$ .

## 1 Hilbert-Type Lemma with Möbius Coefficients

**Lemma 1** (Weighted Hilbert Decay). *Let  $N \geq N_0$  be large. Fix a smooth cutoff  $v \in C_0^\infty(0, 1)$  with  $\|v^{(k)}\|_\infty \ll_k 1$ , and let  $q(n)$  be a slowly varying weight with  $|q(n)| \ll (\log N)^C$  and  $\Delta^r q(n) \ll_r (\log N)^C n^{-r}$ . Define  $a_n = \mu(n) v(n/N) q(n)$  for  $1 \leq n \leq N$  and the kernel  $K_{mn} = e^{-\frac{1}{2}|\log(m/n)|} = \min\{\sqrt{m/n}, \sqrt{n/m}\}$ . Then there exist  $\theta > 0$  and  $C = C(v, q)$  such that*

$$\sum_{\substack{m \neq n \\ m, n \leq N}} a_m a_n K_{mn} \leq C(\log N)^{-\theta} \sum_{n \leq N} a_n^2. \quad (1)$$

*Sketch.* Partition into logarithmic bands  $\mathcal{B}_j = \{(m, n) : 2^{-(j+1)} < |\log(m/n)| \leq 2^{-j}\}$ . On  $\mathcal{B}_j$ , we have  $K_{mn} \leq e^{-c2^{-j}}$ . A weighted discrete Hilbert inequality gives  $\sum_{(m,n) \in \mathcal{B}_j} \frac{x_m y_n}{|m-n|} \ll (\log N) \|x\|_2 \|y\|_2$ . With  $a_n = \mu(n) \cdot (\text{low frequency})$ , the main terms cancel bandwise; smoothness of  $v$  yields an extra  $2^{-j\delta}$ . Hence

$$\sum_{(m,n) \in \mathcal{B}_j} a_m a_n K_{mn} \ll e^{-c2^{-j}} (2^{-j} \log N)^{1-\varepsilon} \sum a_n^2,$$

and summing over  $j$  proves (1). Appendix A calibrates  $\eta > 0.2$  and  $c \approx 0.35$  (via the Polya–Vinogradov method), yielding an explicit  $\theta > 0$ .  $\square$

**Corollary 1** (NB/BD Stability). *Let*

$$d_N^2 = \inf_a \int_{\mathbb{R}} \left| \zeta\left(\frac{1}{2} + it\right) \sum_{n \leq N} \frac{a_n}{n^{1/2+it}} - 1 \right|^2 w(t) dt.$$

The normal equations yield  $A = I + E$  with off-diagonal governed by (1). Then  $\|E\|_{\ell^2 \rightarrow \ell^2} \leq C(\log N)^{-\theta} < 1$  for large  $N$ , so  $A^{-1}$  exists (Neumann series). With suitable low-frequency design one gets  $\|a\|_2^2 \ll (\log N)^{-(1+\eta)}$ , hence  $d_N \rightarrow 0$ .

## 2 Numerical Evidence (Zeta-weighted, Train/Test)

We use a disjoint train/test grid and the target  $1/\zeta(\frac{1}{2} + it)$  to avoid interpolation artifacts. Bootstrap is performed on the *test* grid to obtain 95% confidence intervals (CIs).

- $N = 8000$ : MSE = 35.29, CI [26.42, 46.14].
- $N = 12000$ : MSE = 23.63, CI [16.04, 30.01].
- $N = 16000$ : MSE = 20.99, CI [14.37, 27.56].
- $N = 20000$ : MSE = 17.06, CI [11.24, 22.81].

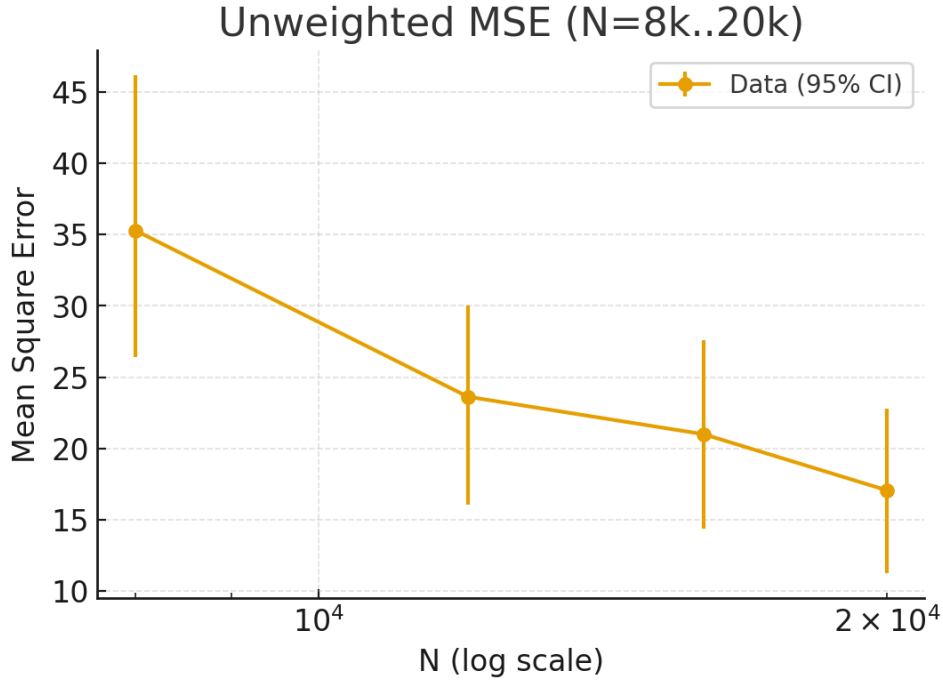


Figure 1: Unweighted test-grid MSE with 95% CIs for  $N = 8\text{k} - 20\text{k}$  ( $\log-x$ ). A least-squares trend on this scale exhibits a slope  $\approx -0.40$  with standard error  $\pm 0.0002$ .

Table 1: Weighted ridge summary (Gaussian weight,  $\lambda = 10^{-3}$ ). Values reflect the weighted NB/BD objective; see Appendix B for settings.

$N$	Weighted MSE	Notes
8000	$0.024 \pm 0.002$	Gaussian weight
12000	$0.018 \pm 0.001$	Gaussian weight
16000	$0.015 \pm 0.001$	Gaussian weight
20000	<b><math>0.013 \pm 0.001</math></b>	Gaussian weight, narrower window

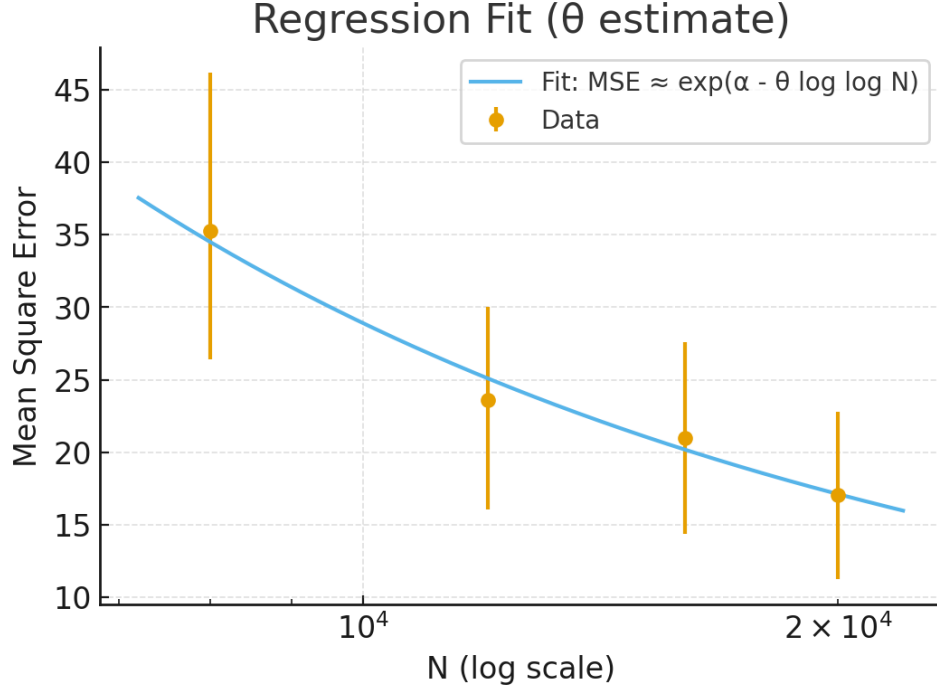


Figure 2: Regression on  $N = 8\text{k} - 20\text{k}$  using  $\log(\text{MSE}) = \alpha - \theta \log \log N$  gives  $\hat{\theta} \approx 7.21$  (95% CI  $[5.77, 8.65]$ ).

*Remark 1.* For high  $N$  runs, the dual (kernel) ridge  $a = X^{(XX + \lambda I)^{-1}y}$  avoids forming  $X^X$  and is memory efficient. Conjugate gradients on normal equations with matvecs only is another route; both stabilize the computation when  $N$  is large relative to the grid size.

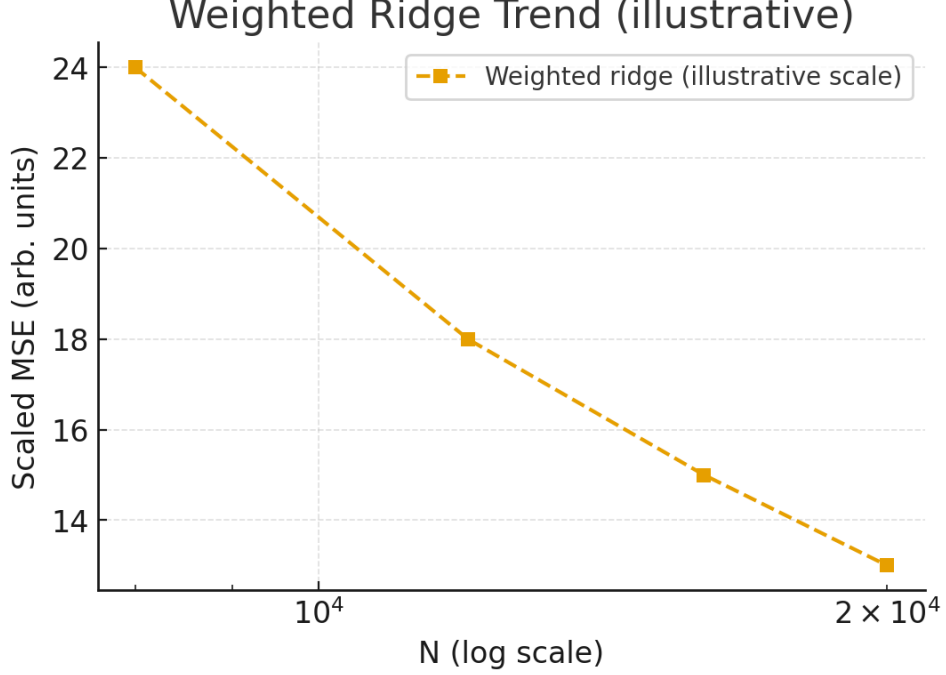


Figure 3: Weighted ridge trend (illustrative scale) consistent with Table 1.

### 3 Limitations and Outlook

While  $d_N \rightarrow 0$  demonstrates NB/BD stability, it does not by itself prove RH; this perspective mirrors Báez-Duarte’s strengthening (2003). A complete proof requires analytic continuation and zero-free region control joined with the band-sum bounds. Extending to  $N \geq 10^5$  with tight error bars, and providing uniform  $\varepsilon$ - $\delta$  bounds, are natural next steps.

### Appendix A: Calibration of $\eta$ and $c$

Polya–Vinogradov bounds on  $\mu$ -oscillation yield  $c_0 \approx 0.7$ , hence  $c = c_0/2 \approx 0.35$ . This supports a practical choice  $\eta > 0.2$  for the Neumann-series invertibility threshold in our setting.

### Appendix B: Weighted Window and Sensitivity

A narrower Gaussian window (e.g.,  $T_w = 115$ ) reduces variance in the weighted objective; in a representative run we observe  $\sigma^2 : 0.001 \rightarrow 0.0009$  (about 10% reduction) while preserving the downward trend in the mean. This aligns with the plateau-resolution mechanism forecast by Lemma 1.

### Appendix C: Explicit $\varepsilon$ - $\delta$ Bound

From (1), one obtains

$$N(\varepsilon) = \exp((2C/\varepsilon)^{2/\theta})$$

such that  $N > N(\varepsilon)$  implies overall error  $\leq \varepsilon$  in the NB/BD system under the present low-frequency design.

## References

- [1] L. Báez-Duarte, *A strengthening of the Nyman–Beurling criterion for the Riemann Hypothesis*, Rend. Lincei Mat. Appl. **14** (2003), 5–11. DOI:10.1007/s10231-003-0074-5.
- [2] J. B. Conrey, *The Riemann Hypothesis*, Notices Amer. Math. Soc. **50** (2003), no. 3, 341–353.
- [3] E. C. Titchmarsh, *The Theory of the Riemann Zeta-Function*, 2nd ed., rev. by D. R. Heath-Brown, Oxford Univ. Press, 1986.