

# Best and near-best polynomial approximation I

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Let  $P_n$  = space of polynomials of degree  $\leq n$ .

Let  $f: [-1, 1] \rightarrow \mathbb{C}$  be a continuous function.

AIM: Replace  $f$  by a proxy  $p_n \in P_n$ . [Do further computation with  $p_n$  as a replacement for  $f$ .]

Many different ways to approximation  $f$ :

• Best  $L_\infty$ ,  $p_n^{L_\infty}$ ,  $p_n^{L_\infty} = \operatorname{argmin}_{q \in P_n} \|f - q\|_\infty$ ,  $\|f\|_\infty = \max_{x \in [-1, 1]} |f(x)|$ .

• Best  $L_2$ ,  $p_n^{L_2}$ ,  $p_n^{L_2} = \operatorname{argmin}_{q \in P_n} \|f - q\|_2$ ,  $\|f\|_2^2 = \int_{-1}^1 |f(x)|^2 dx$ .

• Best  $L_1$ ,  $p_n^{L_1}$ ,  $p_n^{L_1} = \operatorname{argmin}_{q \in P_n} \|f - q\|_1$ ,  $\|f\|_1 = \int_{-1}^1 |f(x)| dx$ .

• ChebT interp,  $p_n^{\text{chebT}}$ ,  $p_n^{\text{chebT}}(x_i) = f(x_i)$ ,  $0 \leq i \leq n$ ,  $x_i = \cos\left(\frac{\pi(i + \frac{1}{2})}{n+1}\right)$ .

• Legendre interp,  $p_n^{\text{leg}}$ ,  $p_n^{\text{leg}}(x_i) = f(x_i)$ ,  $0 \leq i \leq n$ ,  $p_{n+1}(x_i) = 0$ .

• ChebU interp,  $p_n^{\text{chebU}}$ ,  $p_n^{\text{chebU}}(x_i) = f(x_i)$ ,  $0 \leq i \leq n$ ,  $x_i = \cos\left(\frac{i+1}{n+2} \pi\right)$ .

• Leja interp.,  $p_n^{\text{leja}}$ ,  $p_n^{\text{leja}}(x_i) = f(x_i)$ ,  $0 \leq i \leq n$ ,  $x_0, \dots, x_n$  Leja pts

• GE interp.,  $p_n^{\text{GE}}$ ,  $p_n^{\text{GE}}(x_i) = f(x_i)$ ,  $0 \leq i \leq n$ ,  $x_0, \dots, x_n$  chosen greedily.

First observation:  $p_n^{L_1}$ ,  $p_n^{L_2}$ ,  $p_n^{L_\infty}$  are similar:

$$\|f - p_n^{L_1}\|_1 \leq \|f - p_n^{L_2}\|_1 \leq \sqrt{2} \|f - p_n^{L_2}\|_2 \leq \sqrt{2} \|f - p_n^{L_\infty}\|_2 \leq 2 \|f - p_n^{L_\infty}\|_\infty$$

If  $f$  = piecewise polynomial of degree  $\leq n$ :

$$\|f - p_n^{L_\infty}\|_\infty \leq \|f - p_n^{L_2}\|_\infty \leq \sqrt{n} \|f - p_n^{L_2}\|_2 \leq \sqrt{n} \|f - p_n^{L_1}\|_2 \leq \sqrt{n} n^2 \|f - p_n^{L_1}\|_1$$

$C_1$

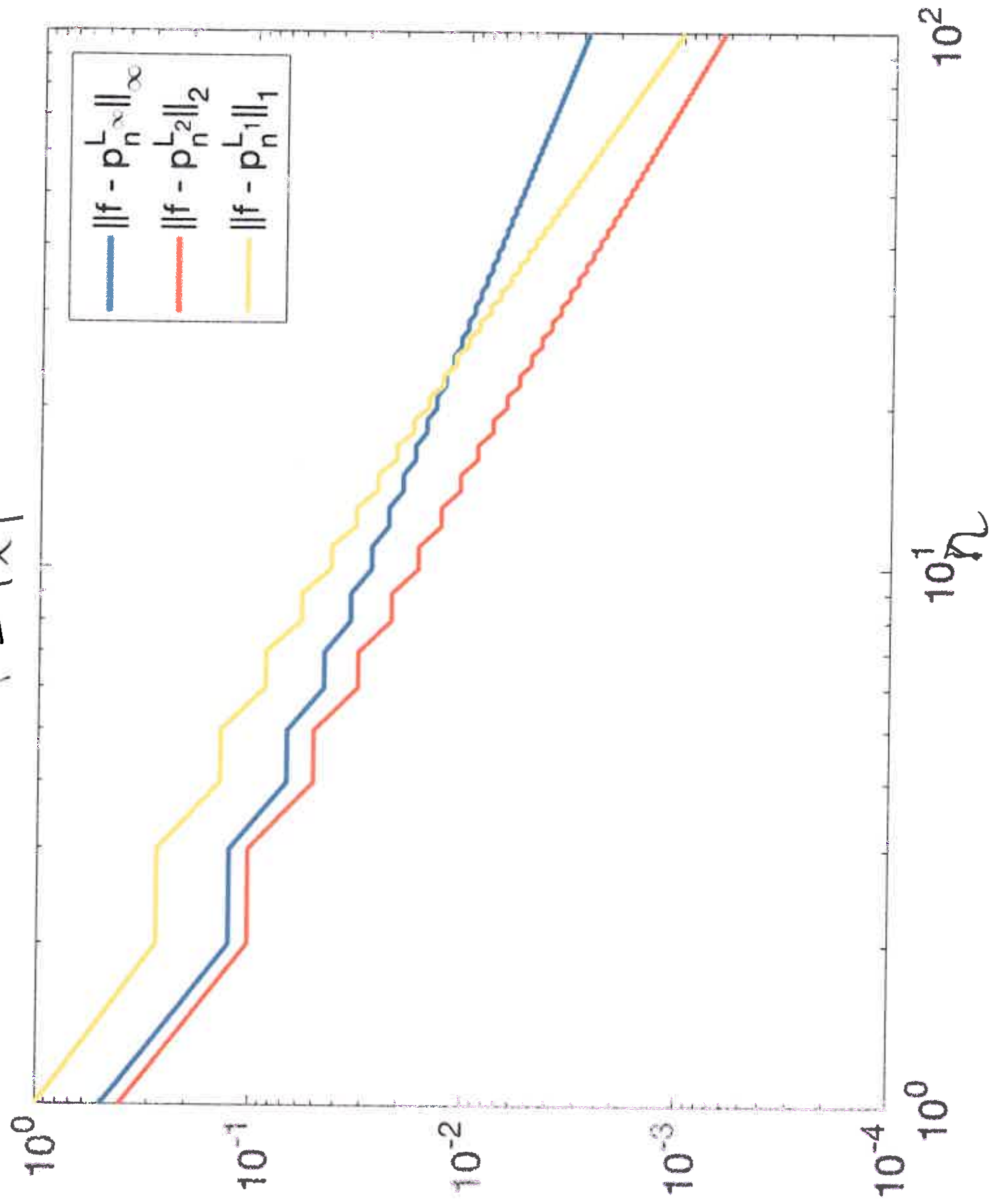
$C_1$

$C_2$

Nikolskii inequalities.

\*

$$F = |x|$$



If  $\|f - p_n^{L\infty}\|_\infty$  decays geometrically, then so does  $\|f - p_n^{L_t}\|_\infty$  for  $t=1, 2, \dots$ .

The main differences are revealed when  $f$  is not particularly smooth.

Best  $L_\infty$  approximation:

Equioscillation (certificate for optimality).

$$p_n^{L\infty} \text{ is best} \iff (f - p_n^{L\infty})(\xi_i) = \pm (-1)^i \|f - p_n^{L\infty}\|_\infty$$

for  $\xi_0, \dots, \xi_{n+1} \in [-1, 1]$

(e.g. Maximum error is attained  $n+2$  in  $[-1, 1]$ .)

Lower bounds on  $\|f - p_n^{L\infty}\|_\infty$ : (De La Vallée - Poussin theorem)

For any  $q \in P_n$ , if  $(f - q)(\xi_i) = -(f - q)(\xi_{i+1})$  for  $\xi_0, \dots, \xi_{n+1} \in [-1, 1]$

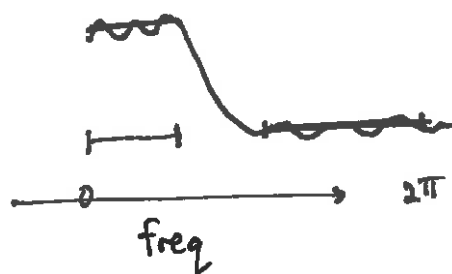
then

$$\min_{0 \leq i \leq n+1} |f(\xi_i) - q(\xi_i)| \leq \|f - p_n^{L\infty}\|_\infty \leq \max_{0 \leq i \leq n+1} |f(\xi_i) - q(\xi_i)|$$

↑  
very important lower bound!

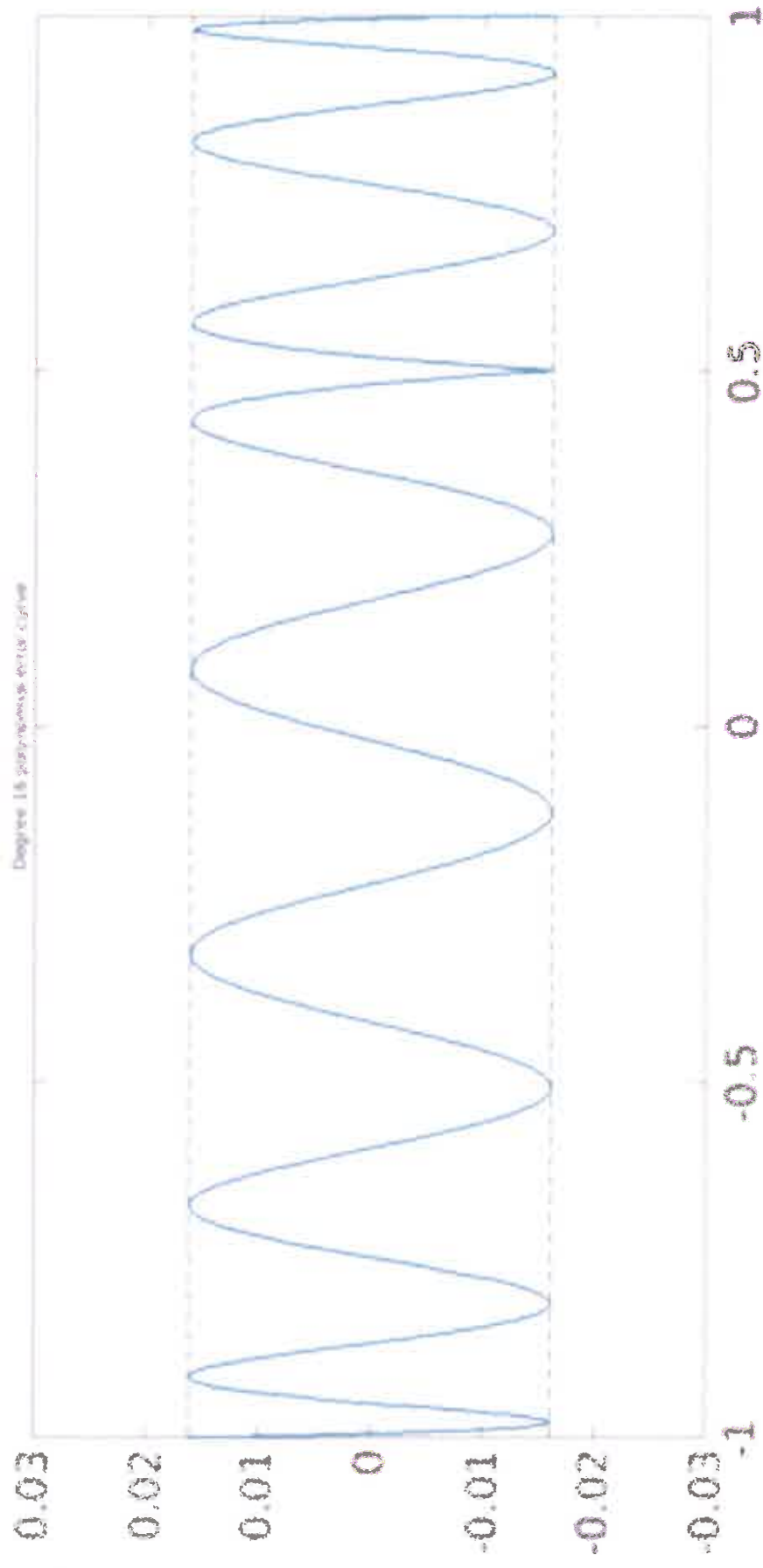
Unfortunately,  $p_n^{L\infty}$  is difficult to compute (done by Remez algorithm) and its applications are restricted to filtering,

where

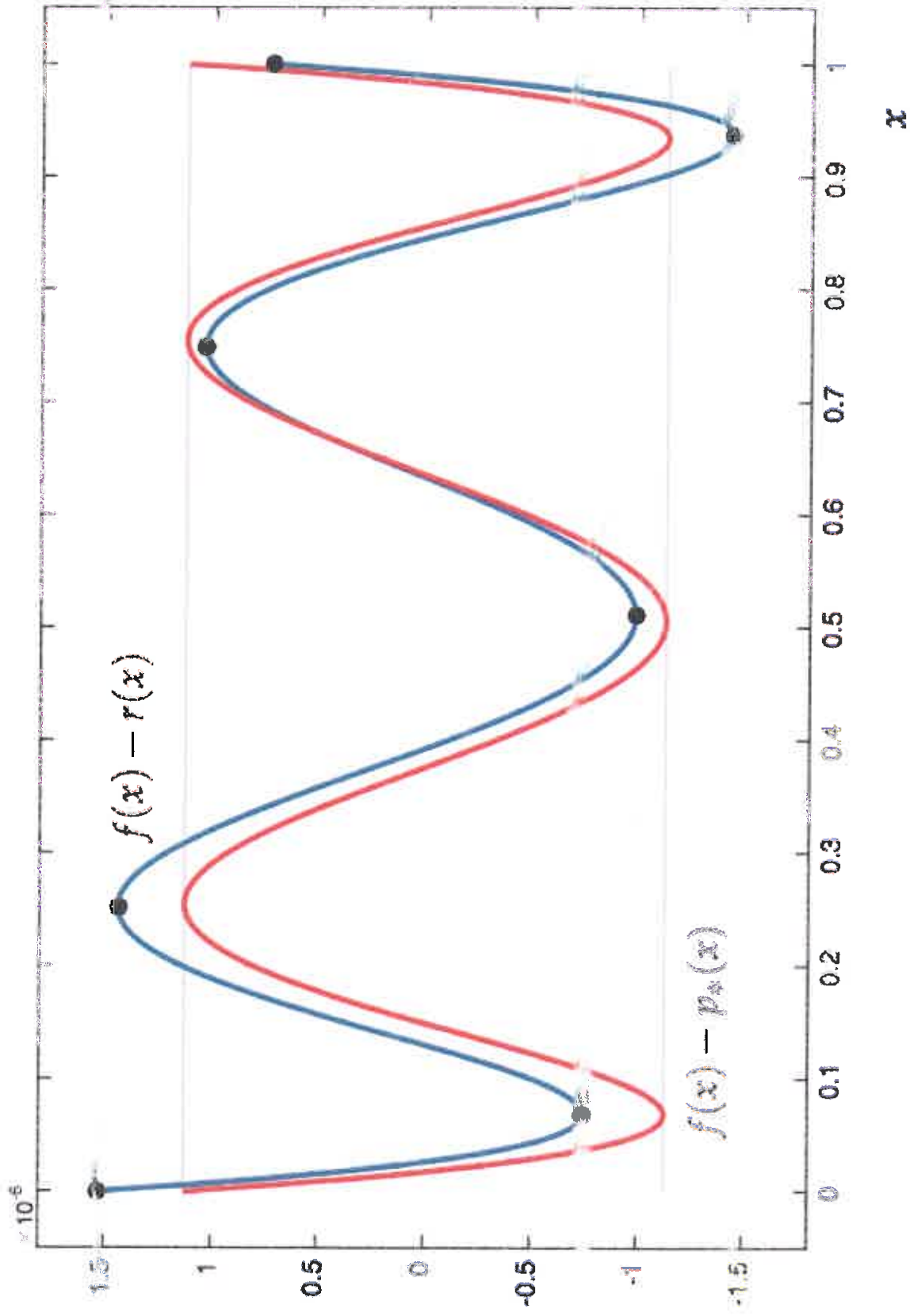


"two intervals".

## Equioscillation theorem.



# De La Vallée - Poussin theorem



Near-best:

$$x^{n+1} - p_n^{\infty}(x) = 2^{-n} T_{n+1}(x), \quad T_{n+1} = \text{chebyshev polynomial.}$$

That means  $p_n^{\infty}(x) = p_n^{\text{chebT}}(x)$  for  $x^{n+1}$ . \*

In fact:  $\|f - p_n^{\text{chebT}}\|_{\infty} \leq (1 + \sigma_n) \|f - p_n^{\infty}\|_{\infty}$

$$\sigma_n \leq 1 + \frac{2}{\pi} \log(n+1)$$

[Erdős, 1962]

The problem: Solve

$$p_n^{\text{chebT}}(x_i) = f(x_i), \quad x_i = \cos\left(\frac{\pi(i + \frac{1}{2})}{n+1}\right)$$

where  $p_n^{\text{chebT}}(x) = \sum_{i=0}^n c_i T_i(x)$

is equivalent to an  $(n+1) \times (n+1)$  linear system:  $A \underline{c} = \underline{f}$ .

The matrix  $A = \text{DCT-II}$  (up to diagonal scaling)

$A^* = \text{DCT-III}$  (up to diagonal scaling).

Therefore,  $A \underline{c} = \underline{f}$  can be solved in  $O(n \log n)$  operation via a FFT [Gentleman, 1970].

Best  $L_2$  approximation:

$$p_n^{L_2} = \underset{p \in P_n}{\operatorname{argmin}} \|f - p\|_2$$

Let  $Q = [\tilde{P}_0(x) | \dots | \tilde{P}_n(x)]$ , where  $\int_{-1}^1 \tilde{P}_i(x) \tilde{P}_j(x) dx = \begin{cases} 1, & i=j \\ 0, & \text{o/w} \end{cases}$ .

and  $\text{degree}(\tilde{P}_i(x)) = i$ . Known as ~~normalized~~ ~~Legendre~~

Legendre polynomials.

Then,  $p = Q \underline{c}$  for all  $p \in P_n$ .

Therefore,  $p_n^{L_2} = \underset{\underline{c} \in \mathbb{C}^{n+1}}{\operatorname{argmin}} \|f - Q \underline{c}\|_2$

This is a least squares problem.

Solution:  $Q^T Q \underline{c} = Q^T f \Rightarrow \underline{c} = \underbrace{Q^T f}_{\text{projection of } f \text{ onto } P_n}$

$p_n^{L_2}$  is computed by calculating  $c_k = \int_{-1}^1 f(x) \tilde{P}_k(x) dx$  for  $0 \leq k \leq n$ .

Legendre interpolation: Solve  ~~$\tilde{P}_n \underline{c} = \underline{f}$~~

Solve  $p_n^{\text{leg}}(x_i) = f(x_i)$ ,  $\tilde{P}_{n+1}(x_i) = 0$

$$p_n^{\text{leg}}(x) = \sum_{k=0}^n c_k \tilde{P}_k(x)$$

$\Rightarrow$  Solve:  $\tilde{P}_n \underline{c} = \underline{f}$ ,  $(\tilde{P}_n)_{jk} = \tilde{P}_k(x_j)$

By Gauss quadrature:  $\{x_i, w_i\}$

$$\int_{-1}^1 \tilde{P}_j(x) \tilde{P}_k(x) dx = \sum_{i=0}^n w_i \tilde{P}_j(x_i) \tilde{P}_k(x_i) = \begin{cases} 1, & j=k \\ 0, & j \neq k \end{cases}$$

$$\therefore \tilde{P}_n^T D_w \tilde{P}_n = I$$

$$\tilde{P}_n \underline{c} = \underline{f} \Rightarrow \underline{c} = D_w \tilde{P}_n^T \underline{f} \Rightarrow c_k = \sum_{i=0}^n w_i \tilde{P}_k(x_i) f(x_i)$$

Near-best:

$$\|f - p_n^{\text{leg}}\|_2^2 = \|f - p_n^{L_2}\|_2^2 + \|p_n^{L_2} - p_n^{\text{leg}}\|_2^2 \quad \left( \langle f - p_n^{L_2}, q \rangle = 0 \text{ for any } q \in P_n \right)$$

$$\|p_n^{L_2} - p_n^{\text{leg}}\|_2^2 = \sum_{i=0}^n w_i (p_n^{L_2}(x_i) - f(x_i))^2 \approx \int_{-1}^1 (f - p_n^{L_2})(x) dx$$

↑ Gauss quadrature error

Since  $f$  is continuous:

$$\|f - p_n^{\text{leg}}\|_2 \leq \left(2 + \frac{c}{n}\right)^{\frac{1}{2}} \|f - p_n^{\text{L}}\|_2 \quad [\text{Xiang \& Bornemann 2012}]$$

Since  $c = D_w P_n^T f$ , the polynomial  $p_n^{\text{leg}}$  can be computed in  $O(n \log n)$  operations via a FFT-based transform [Hale \& T. 2015].

### Noisy function approximation

Suppose one only knows  $f: [-1, 1] \rightarrow \mathbb{R}$  up to a noise level  $s^2$ .

That is,  $\hat{f}(x_i) = f(x_i) + \varepsilon_i$   $\varepsilon_i \sim N(0, s^2)$

$$\begin{aligned} \mathbb{E}[\|\hat{c} - c\|_2^2] &= \mathbb{E}[\|D_w P_n^T \varepsilon\|_2^2] \leq \|D_w^{\frac{1}{2}}\|_2^2 \mathbb{E}[\|\varepsilon\|_2^2] \\ &\leq \frac{\pi}{n+1} \cdot (n+1) s^2 = \pi s^2. \end{aligned}$$

Moreover,

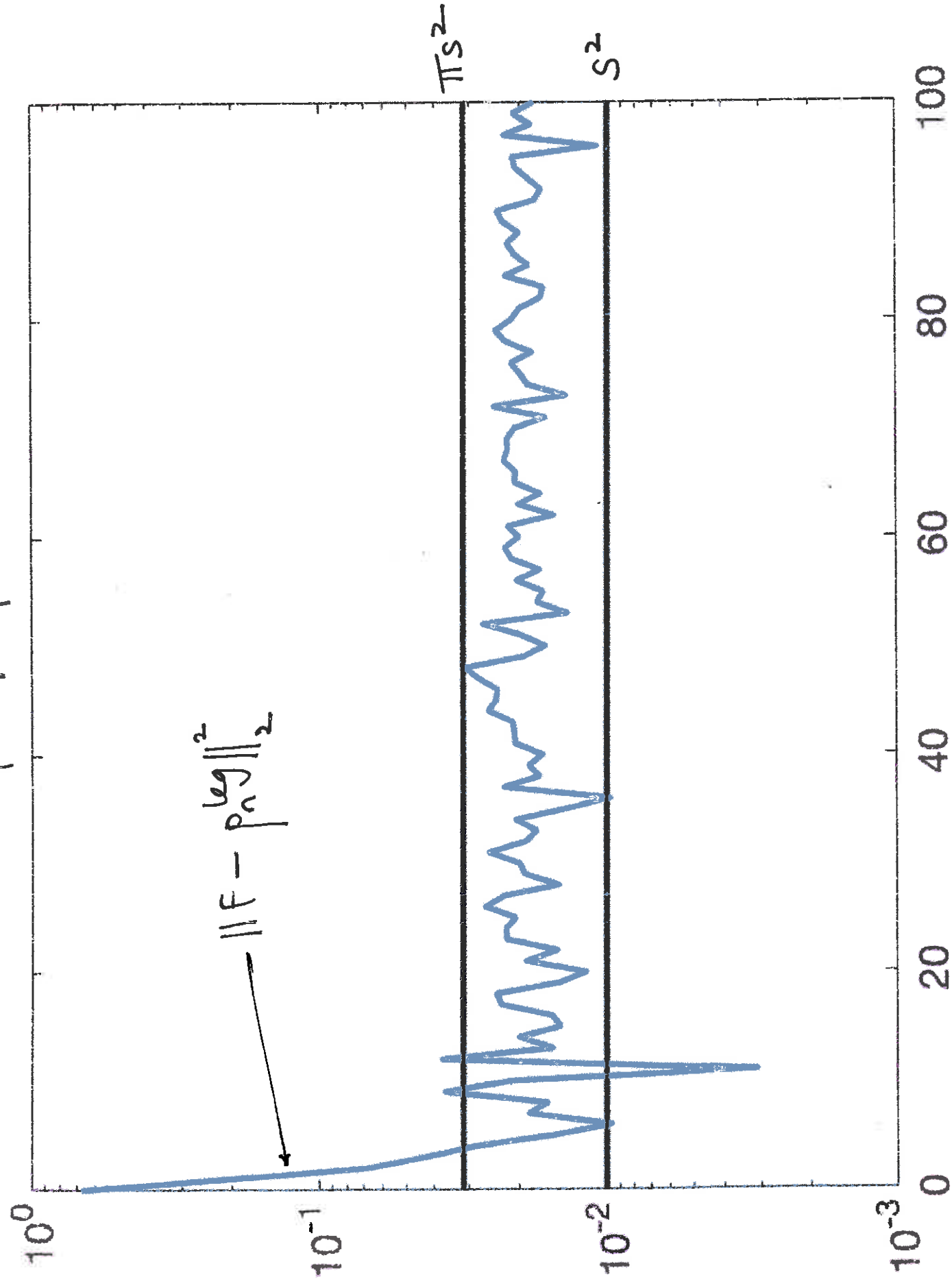
$$\mathbb{E}[\|\hat{p}_n^{\text{leg}} - p_n^{\text{leg}}\|_2^2] = \mathbb{E}[\|\hat{c} - c\|_2^2] \leq \pi s^2.$$

Recovery of noisy  $f$ , up to the noise level. This can be performed/computed in  $O(n \log n)$  operations.

Conclusion: Legendre Interpolation is ideal for approximating noisy functions.



$$f = |x| + \text{noise}$$



## Best and near-best polynomial approximation II

### Best $L_1$ approximation

(Characterization): If  $f(x) - p_n^{L_1}(x)$  has finite many zeros, then

$$p_n^{L_1} \text{ best} \iff \int_{-1}^1 \text{sign}(f(x) - p_n^{L_1}(x)) q(x) dx = 0 \text{ for } \forall q \in P_n$$

While this isn't as visual as equioscillation theorem, it is a remarkable result. For example; it tells us that:

$$|\{x \in [-1, 1] : f(x) - p_n^{L_1}(x) > 0\}| \approx |\{x \in [-1, 1] : f(x) - p_n^{L_1}(x) < 0\}|$$

Corollary: If  $f(x) - p_n^{L_1}(x)$  has exactly  $n+1$  zeros (which it does generically / usually) in  $[-1, 1]$ , then

$$p_n^{L_1}(x) = p_n^{\text{cheb}}(x)$$

Here,  $p_n^{\text{cheb}}(x)$  is the polynomial interpolant of  $f$  at the zeros of  $U_{n+1}(x)$ .  $U_{n+1} =$  the Chebyshev polynomial of the second kind

$$= \frac{\sin((n+2)\cos^{-1}(x))}{\sin(\cos^{-1}(x))}$$

We conclude that  $p_n^{\text{cheb}}(x)$  is very often the best  $L_1$  polynomial approximant. It can be computed via an FFT (more precisely a DST) in  $O(n \log n)$  operations.

For example:  $x^{n+1} - p_n^{L_1}(x) = b_n U_{n+1}(x)$  ( $b_n$  to make  $b_n U_{n+1}$  monic)

Proof:  $U_{n+1}(x)$  has precisely  $n+1$  zeros.

Fiedler and Jurskat (1990) gave a large class of  $f(x)$  such that  $p_n^{L_1}(x) = p_n^{\text{cheb}}(x)$ . Including:  $|x|$ ,  $\sqrt{1-x^2}$ ,  $\frac{1}{x^2+a^2}$ ,  $\sin^{-1}(x)$ , ...

Near-best: I believe that:

$$\|f - p_n^{\text{cheb}}\|_1 \leq (1 + \tau_n) \|f - p_n^{L_1}\|_1$$

$$\tau_n \leq 1 + \frac{2}{\pi} \log(n+2)$$

[A very closely related result in Freilich & Mason 1971.]

## \* Error concentration/localization

Recall that  $\|f - p_n^{L_1}\|_1 \leq 2 \|f - p_n^{L_\infty}\|_\infty \leq 2 C_2 n^2 \|f - p_n^{L_1}\|_1$  if  $f = \text{piecewise poly. of degree } \leq n$ .

Therefore,  $\|f - p_n^{L_1}\|_1$  can converge to zero at a faster rate than  $\|f - p_n^{L_\infty}\|_\infty$ . But,  $\|f - p_n^{L_1}\|_\infty \geq \|f - p_n^{L_\infty}\|_\infty$ .

If  $\|f - p_n^{L_1}\|_1 \ll \|f - p_n^{L_\infty}\|_\infty$ , then the only choice is

for  $f(x) - p_n^{L_1}(x)$  to have localized error.

$$\text{If } \Omega_n = \{x \in [-1, 1] : f(x) - p_n^{L_1}(x) > \frac{1}{2} (f(x) - p_n^{L_\infty}(x))\}$$

then

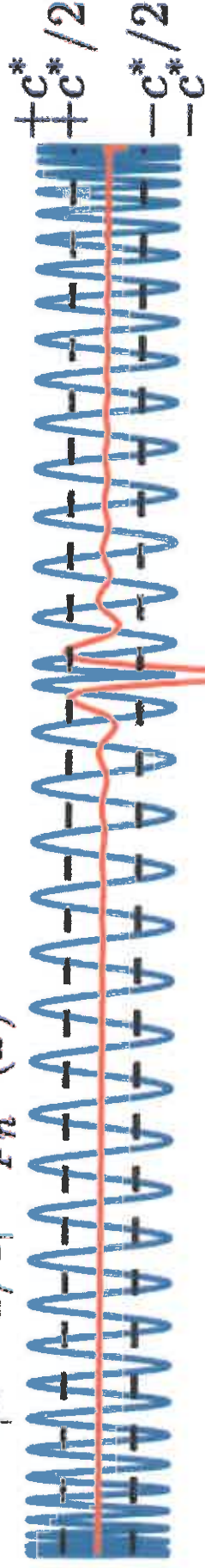
$$|\Omega_n| \leq \frac{2 \|f - p_n^{L_1}\|_1}{\|f - p_n^{L_\infty}\|_\infty}$$

For example, if  $f(x) = |x|$ :

- $\|f - p_n^{L_1}\|_1 \sim \frac{\pi^2}{4n^2}$
- $\|f - p_n^{L_\infty}\|_\infty \sim \frac{0.28017}{2n}$

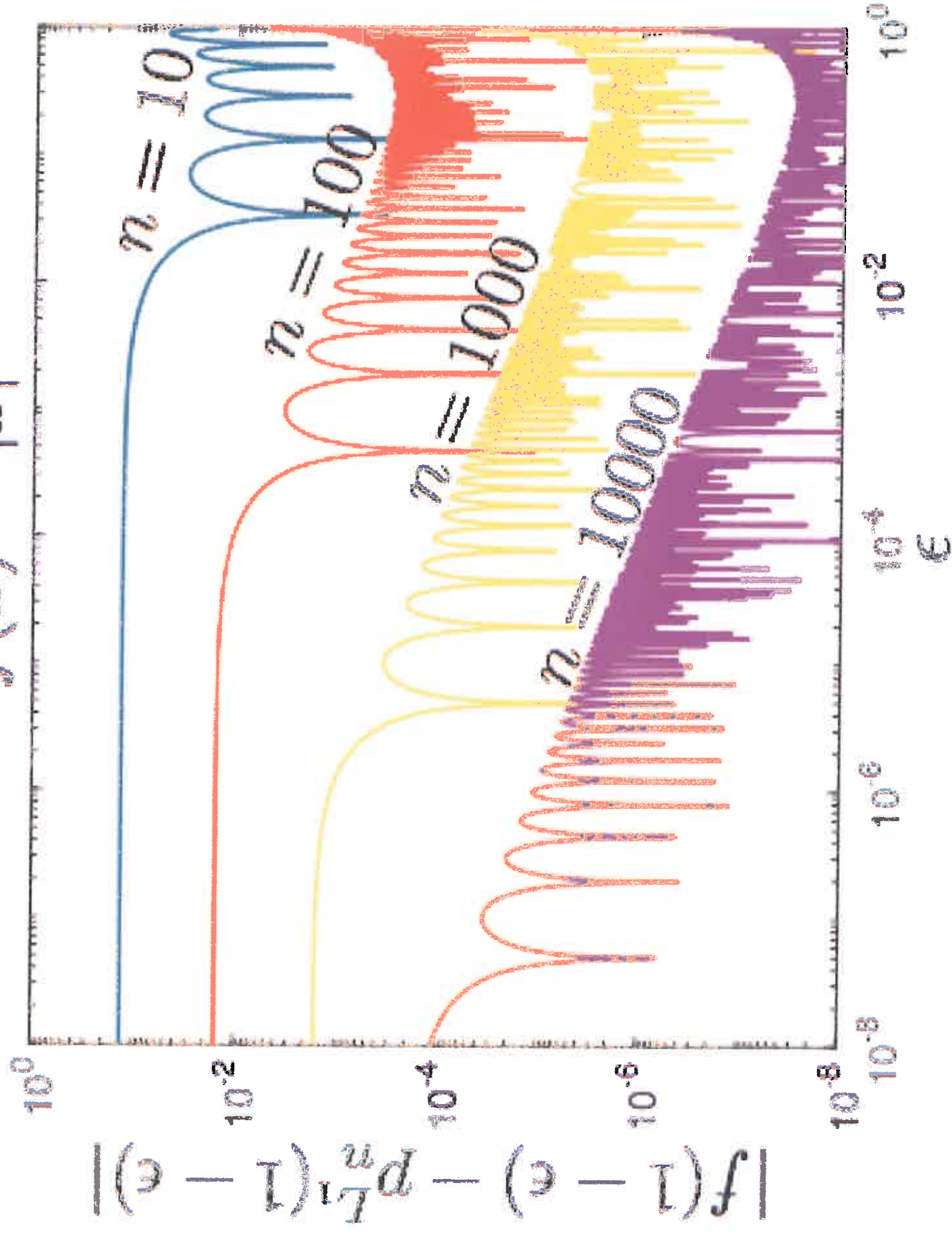
$\Rightarrow |\Omega_n| \sim \sim \frac{0.28017}{2n} \cdot \frac{\pi^2}{4n^2} = O\left(\frac{1}{n}\right)$ .

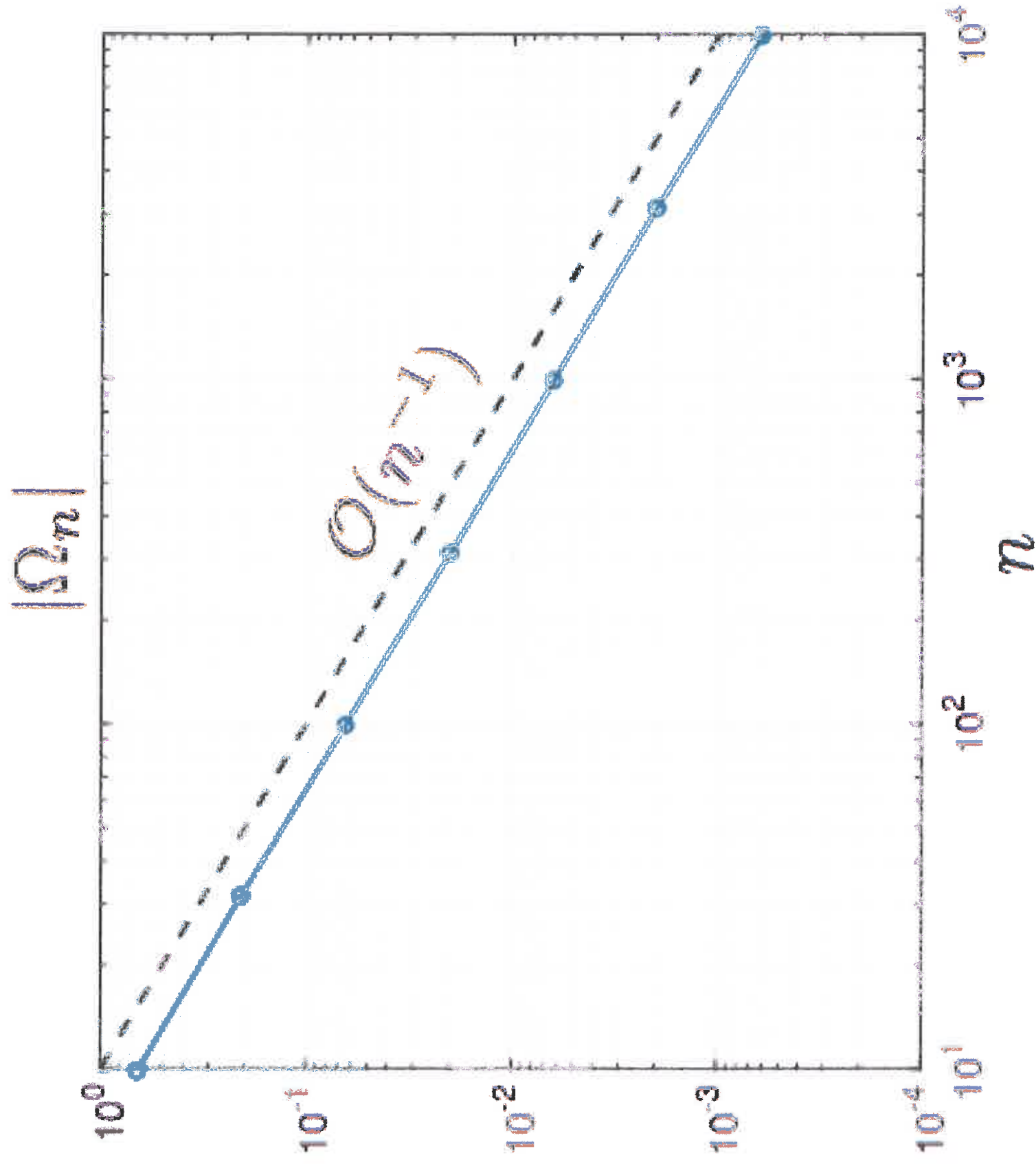
$$|x - 1/4| - p_n^{L_\infty}(x)$$



$$|x - 1/4| - p_n^{L_1}(x)$$

$$f(x) = |x|$$





The error localization properties make  $p_n^{L1}$  ideal for employing on functions that have been corrupted on small intervals.

~~Leja~~ int

GE on functions: If a function is very expensive to evaluate, then  $p_n^{L\infty}$ ,  $p_n^{L2}$ ,  $p_n^{L1}$  or their faster variants  $p_n^{\text{chebt}}$ ,  $p_n^{\text{leg}}$ ,  $p_n^{\text{chebu}}$  may not be suitable:

- Requires lots of function evaluation to compute,
- If, for a fixed  $n$ , one has not resolved  $f$  sufficiently, then ~~there is also~~  $p_{n+1}$  costs another  $n+1$  function evaluation.

Gaussian elimination:

- ① Set  $p(x) = 0$ . Set  $\underline{x} = \{ \}$  empty set.
- ① Select  $x_* = \arg \max |f(x) - p(x)|$
- ② ~~Select~~ <sup>update</sup>  $\underline{x} \leftarrow \{ \underline{x}, x_* \}$  and set  $p(x)$  to be the polynomial interpolant of  $f$  at  $\underline{x}$ .
- ③ Repeat ① & ②.

Conjecture / observation:

For reasonable functions  $f$ , the point set  $\underline{x}$  is distributed (in some sense) like the Chebyshev points.

Convergence: All that is known: If  $f: [-1, 1] \rightarrow \mathbb{R}$  is analytic in a stadium of radius  $\beta > 2$  and bounded on its closure, then

$$\|f - p_n^{\text{GE}}\|_{\infty} \leq C \left( \frac{2}{\beta} \right)^{-n} \quad \text{i.e. geometrically fast.}$$

