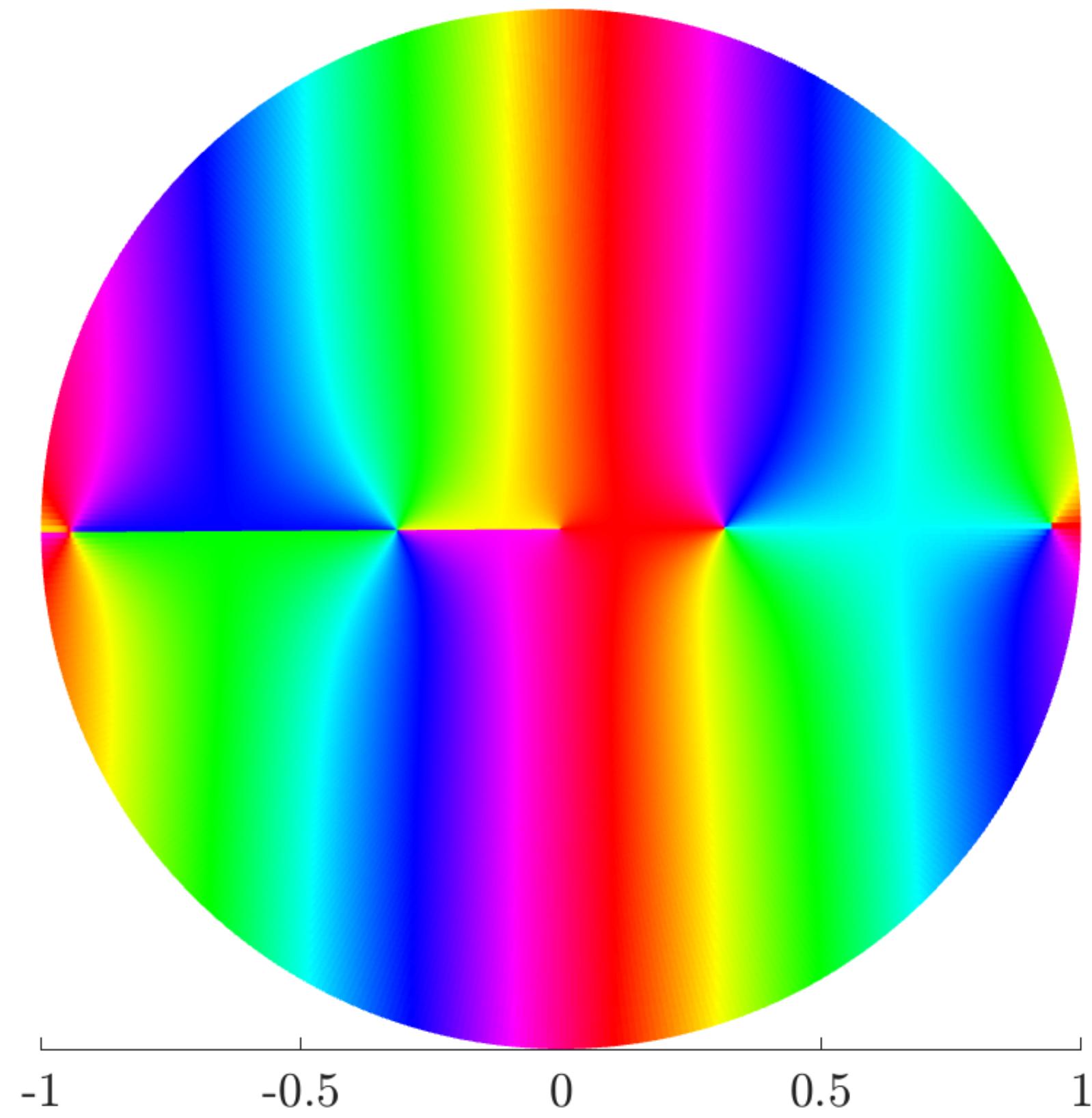
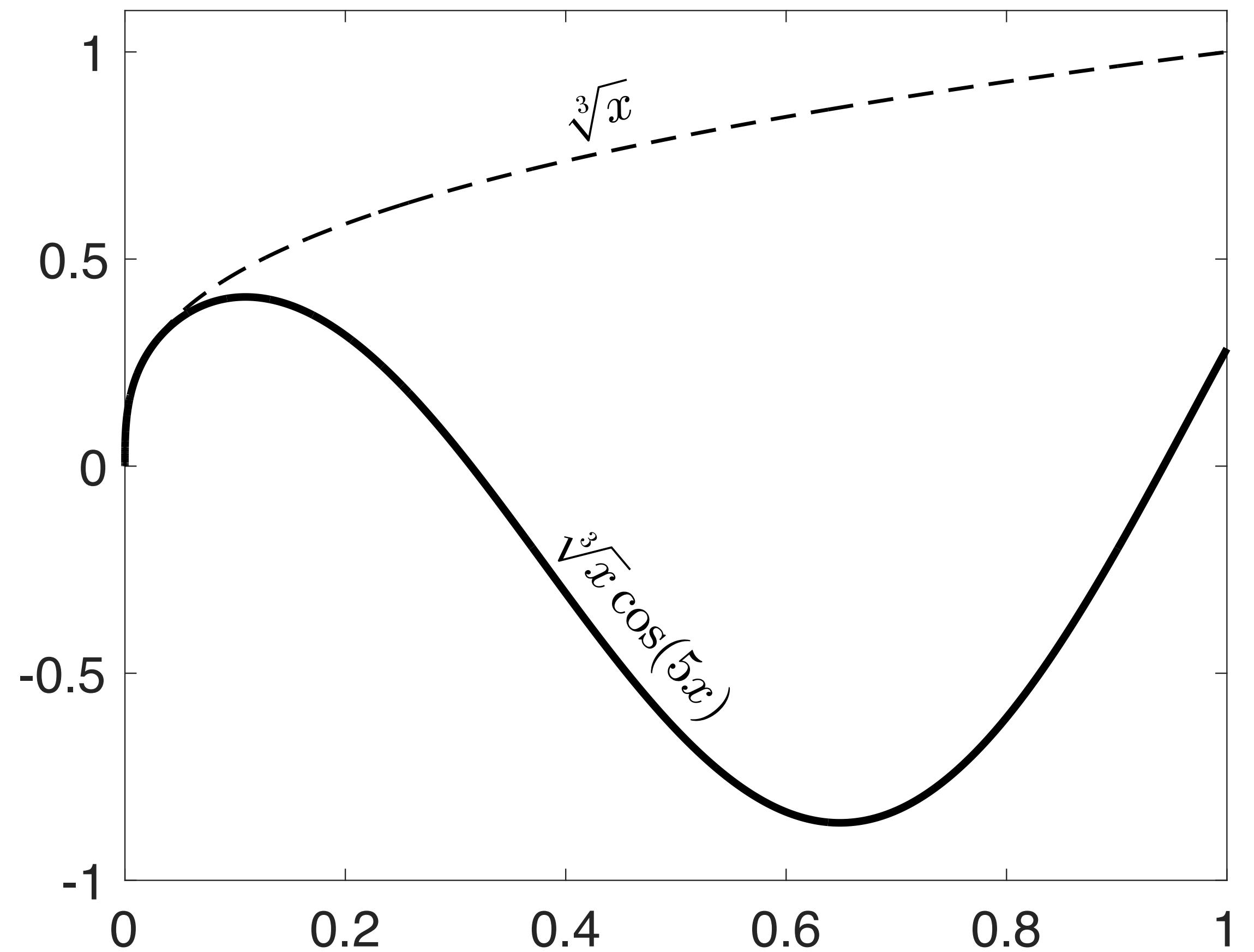


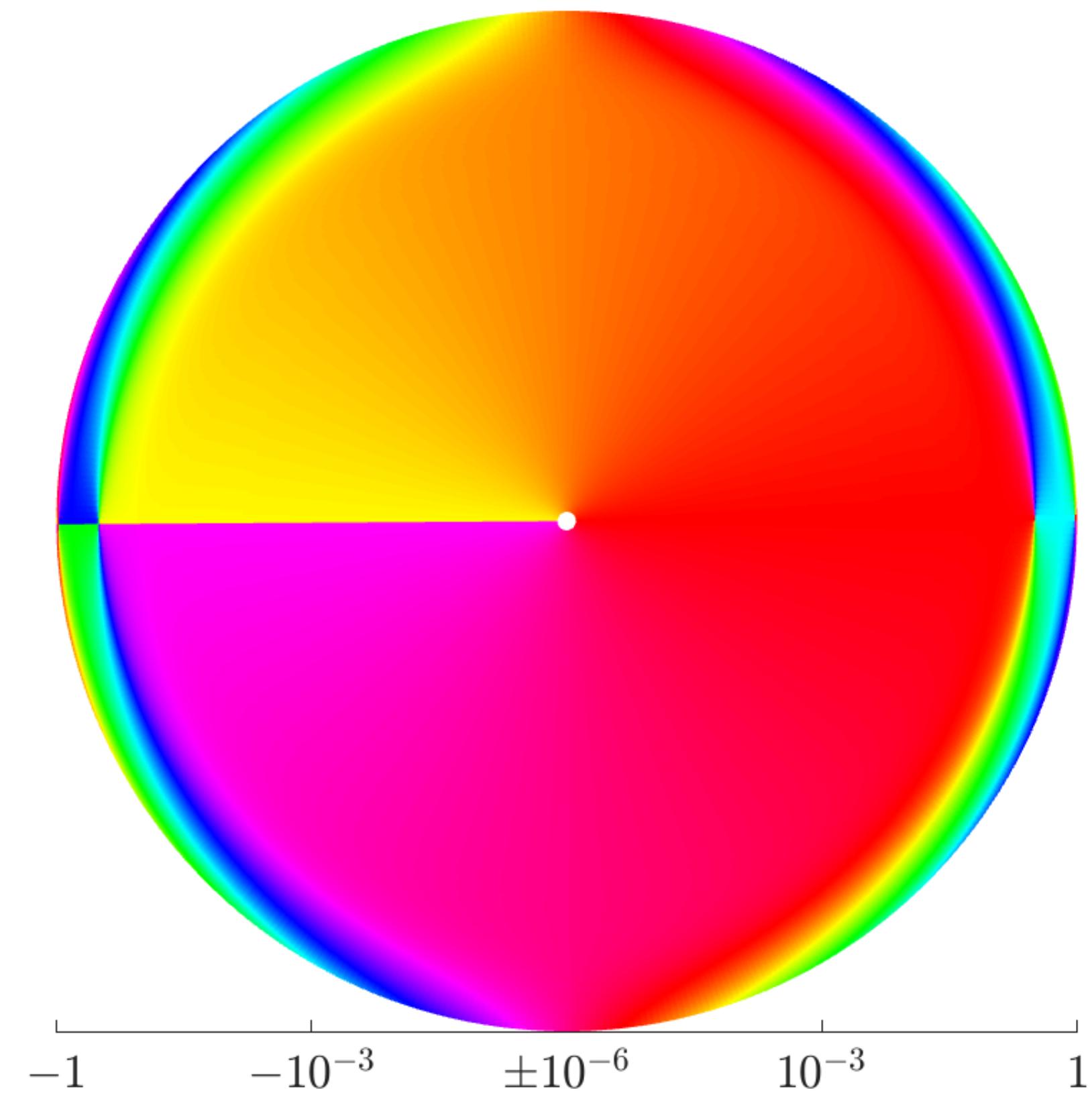
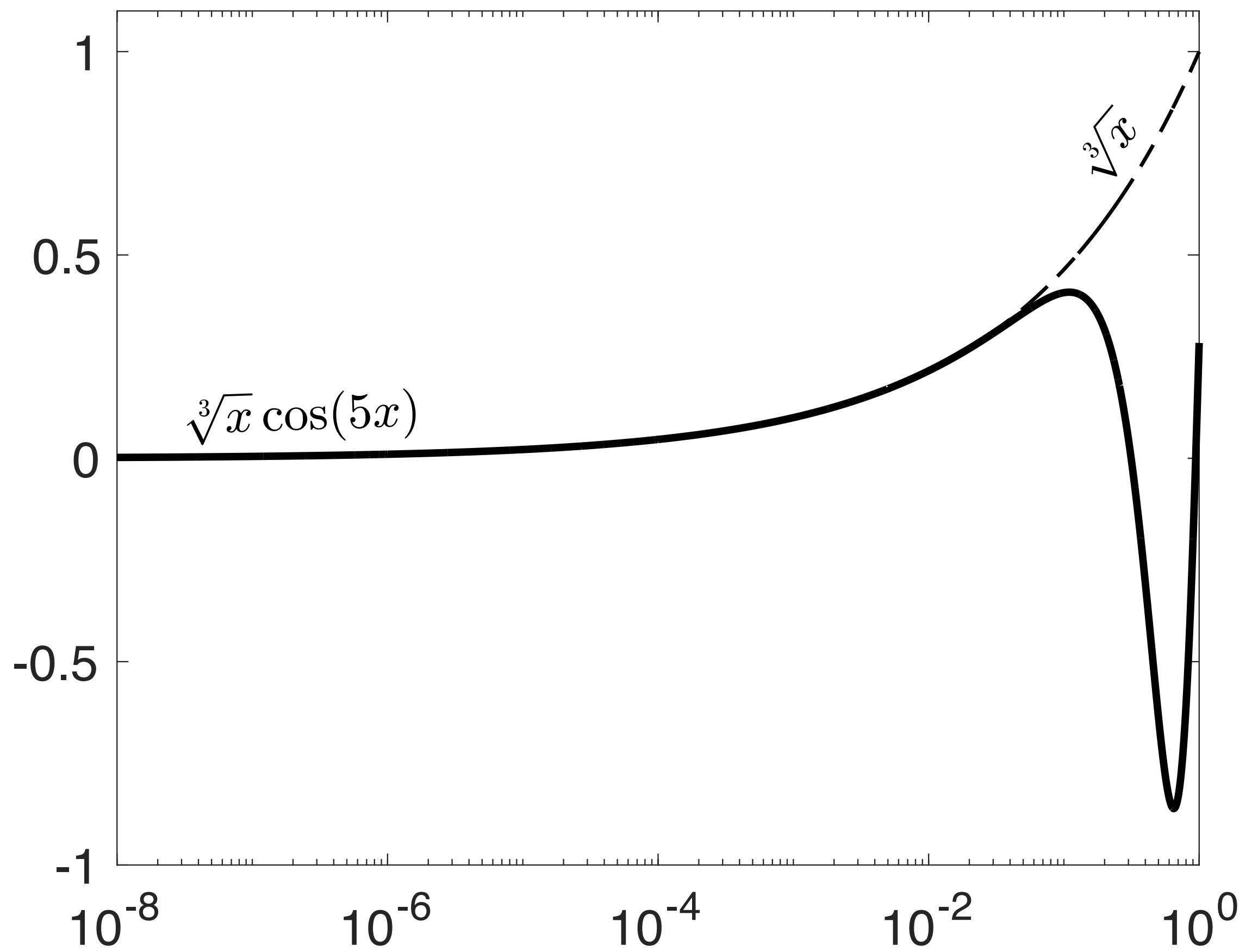
Rational approximation using partial fractions with preassigned poles

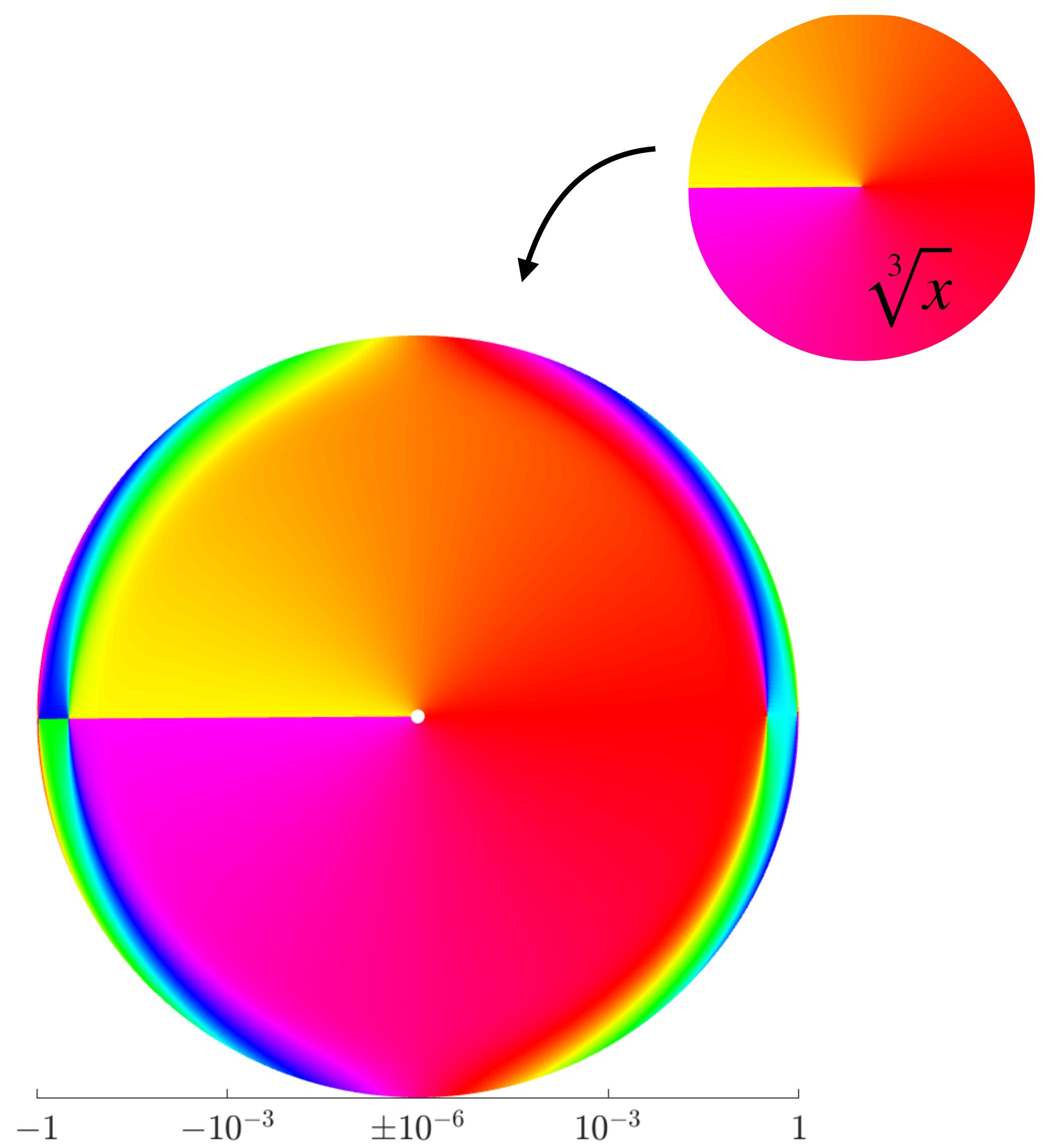
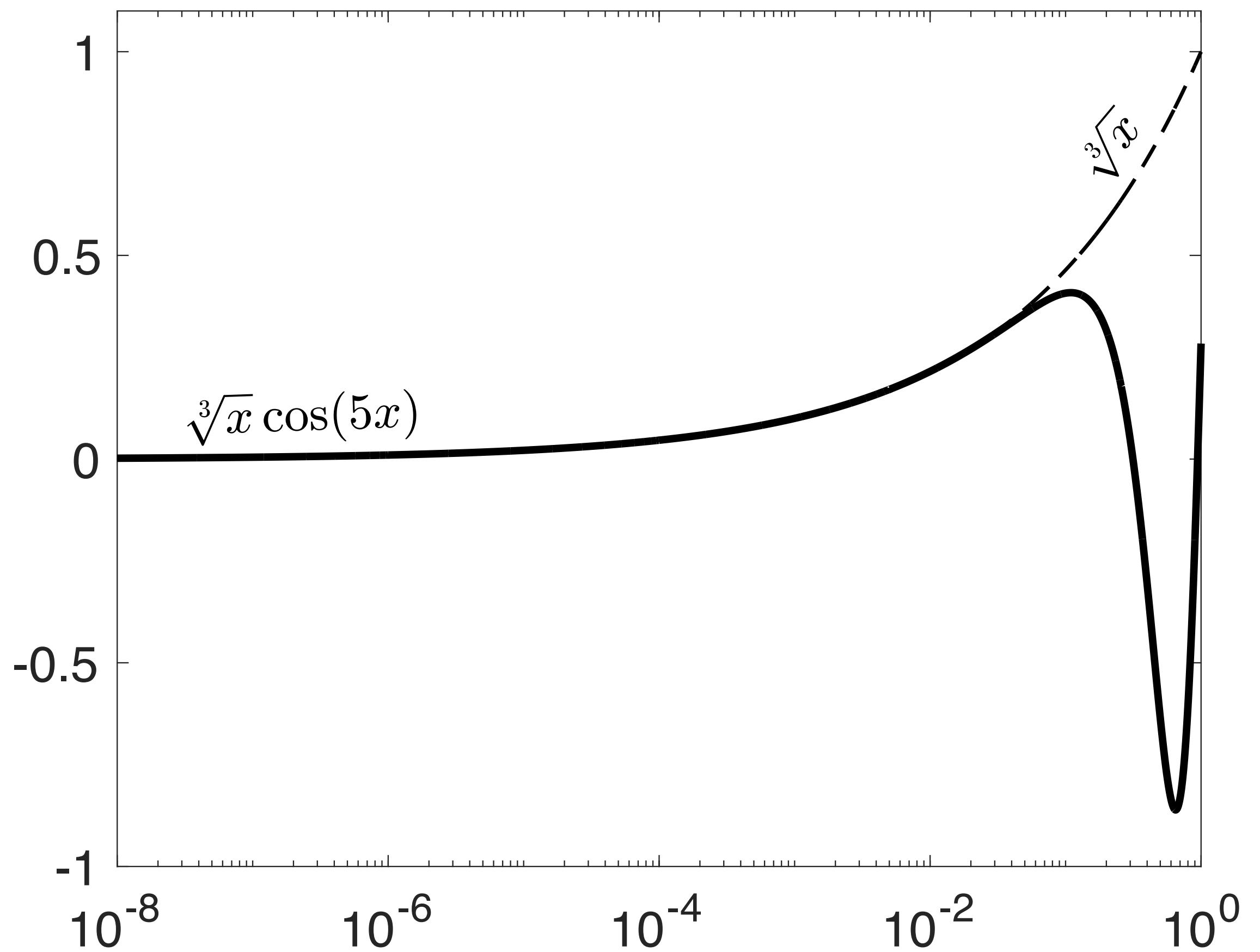
Astrid Herremans

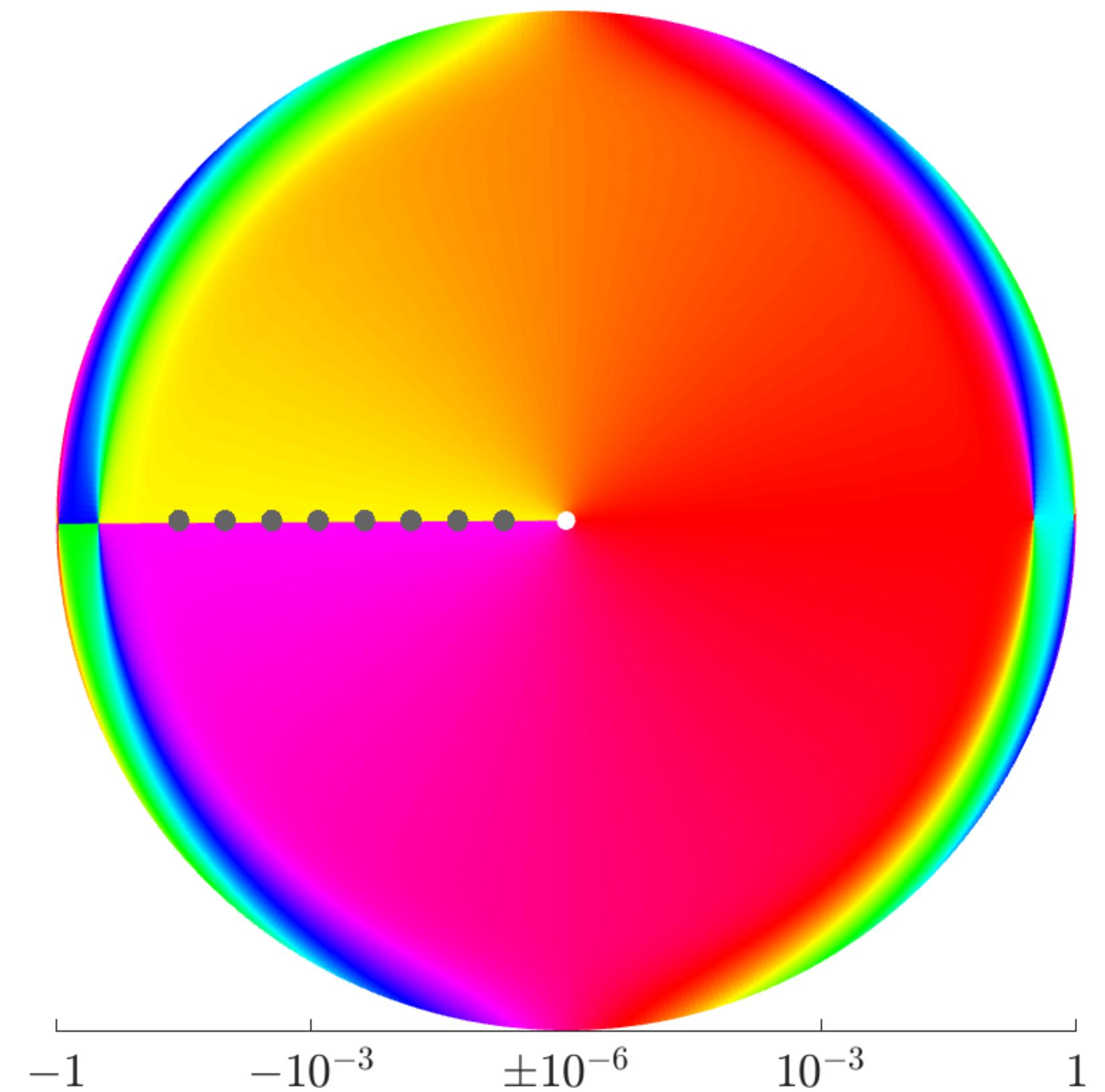
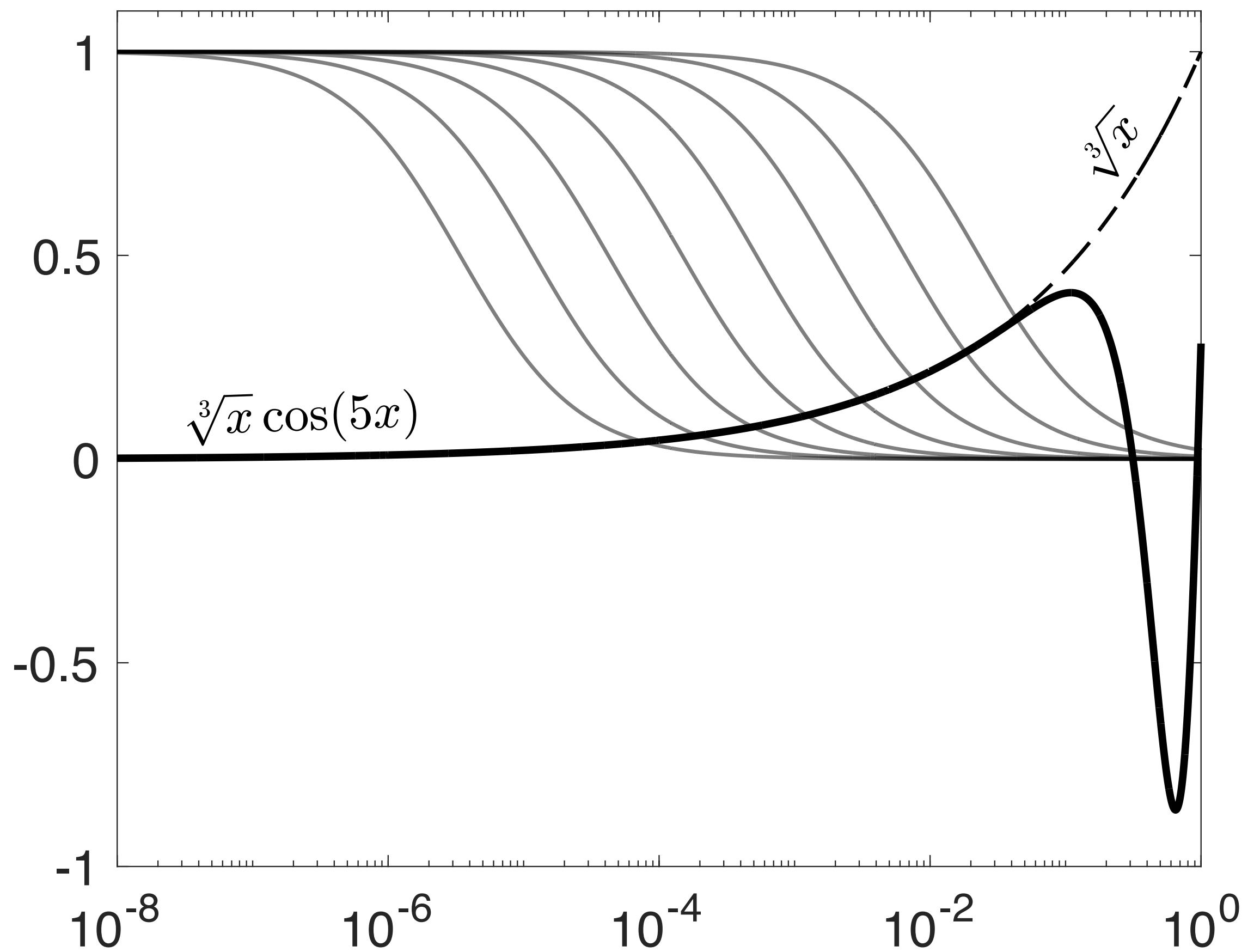
joint work with Daan Huybrechs, Nick Trefethen and Nicolas Boullé

- ▶ the univariate case
- ▶ the multivariate case
- ▶ on the partial fractions representation







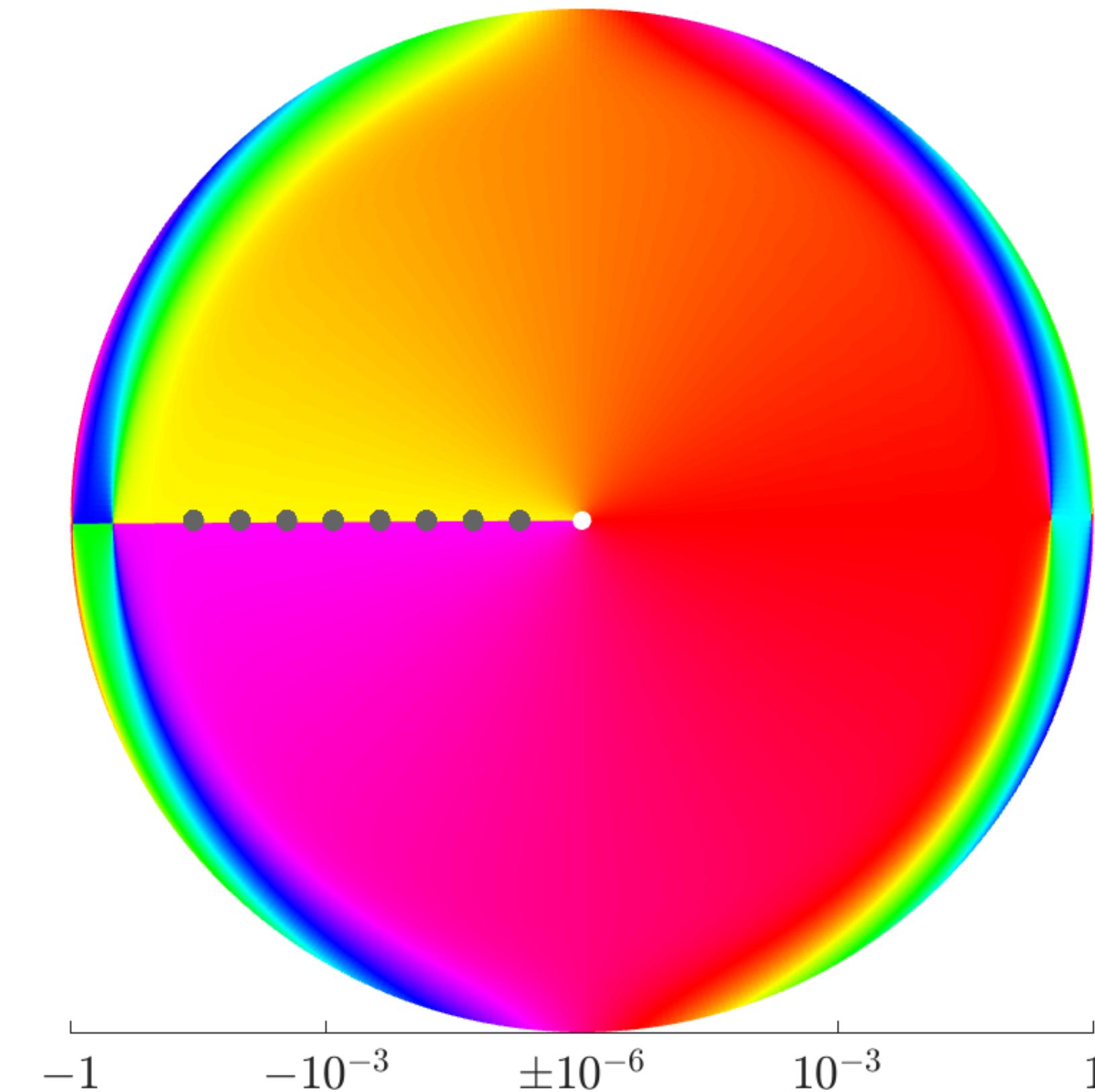
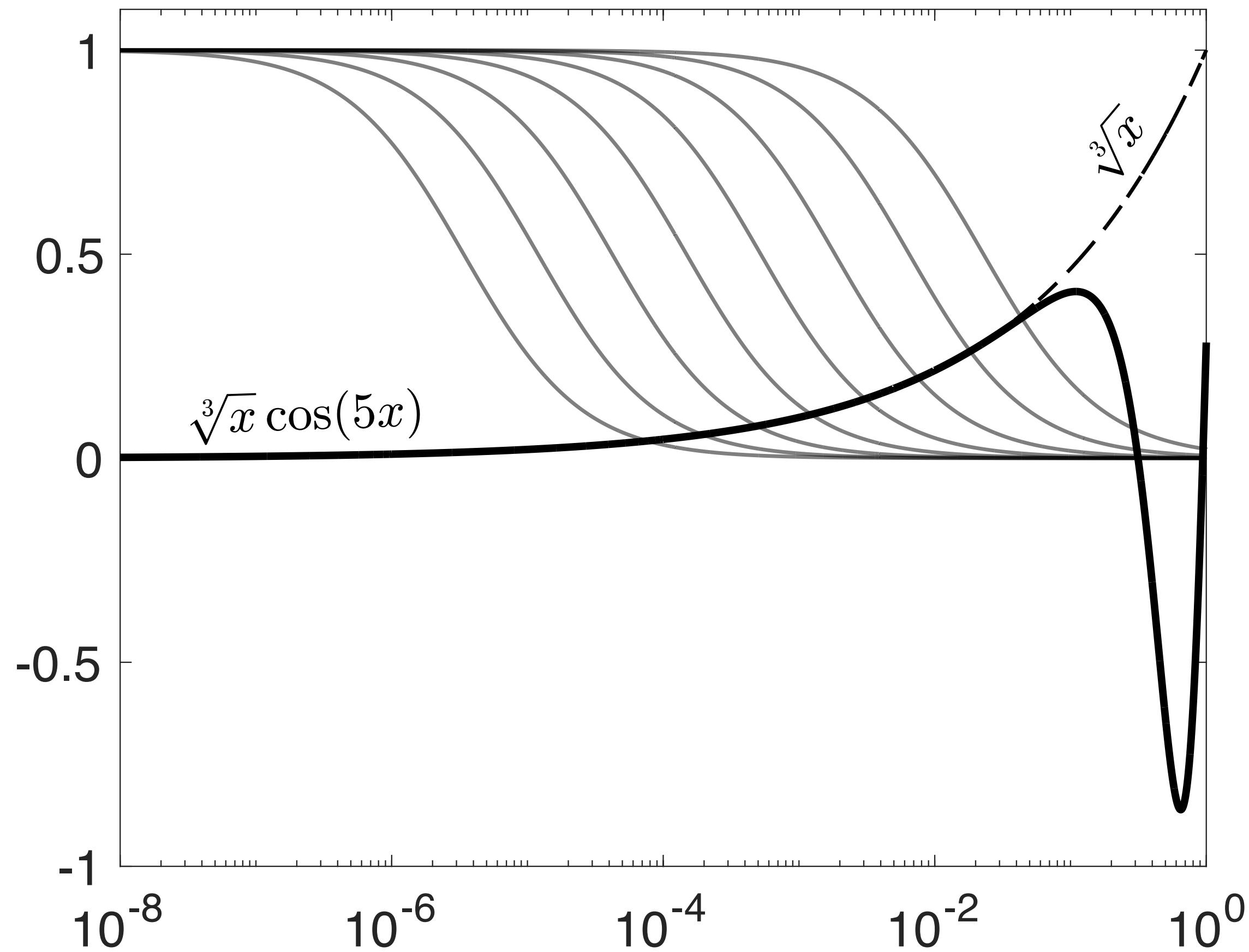


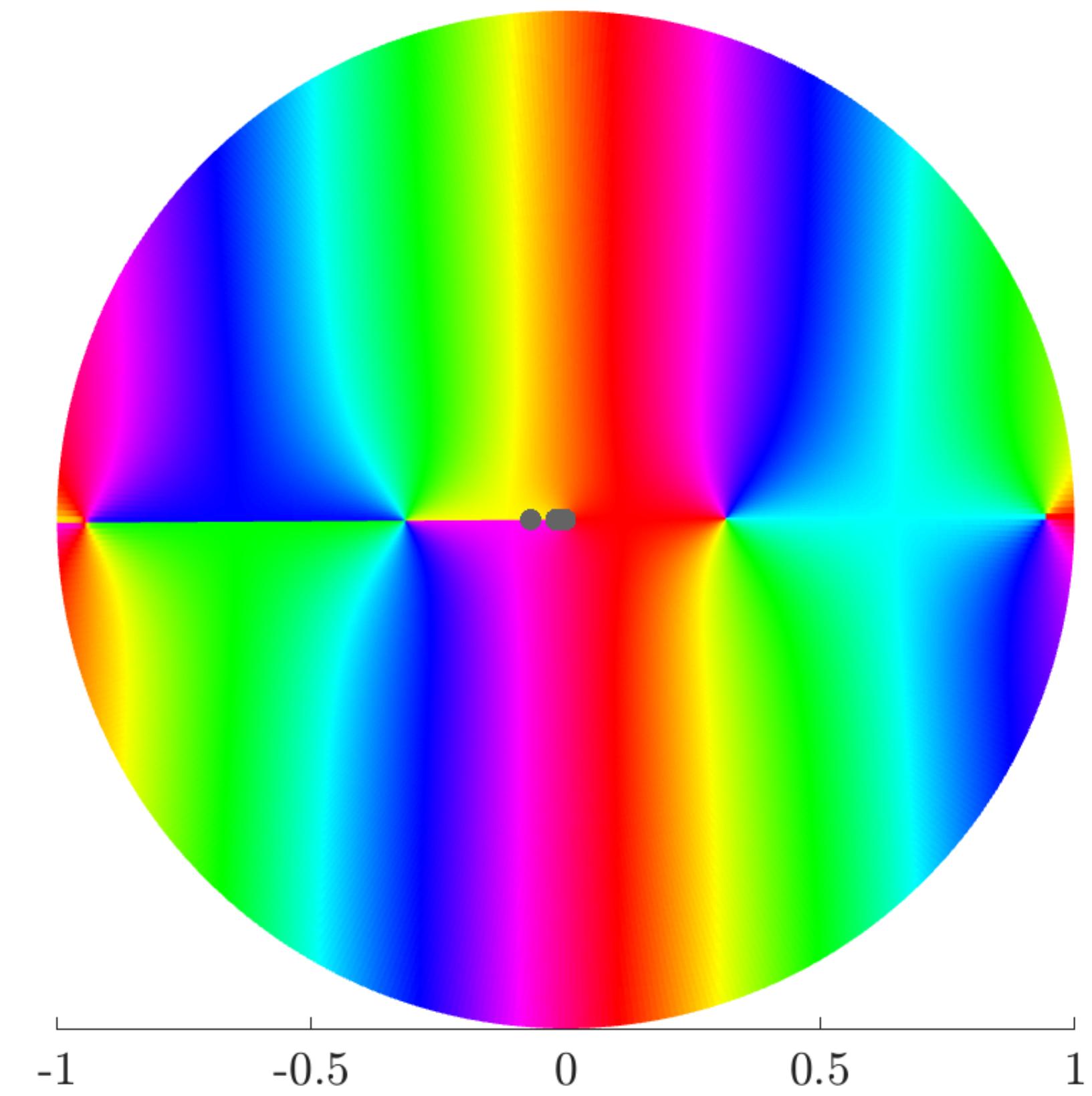
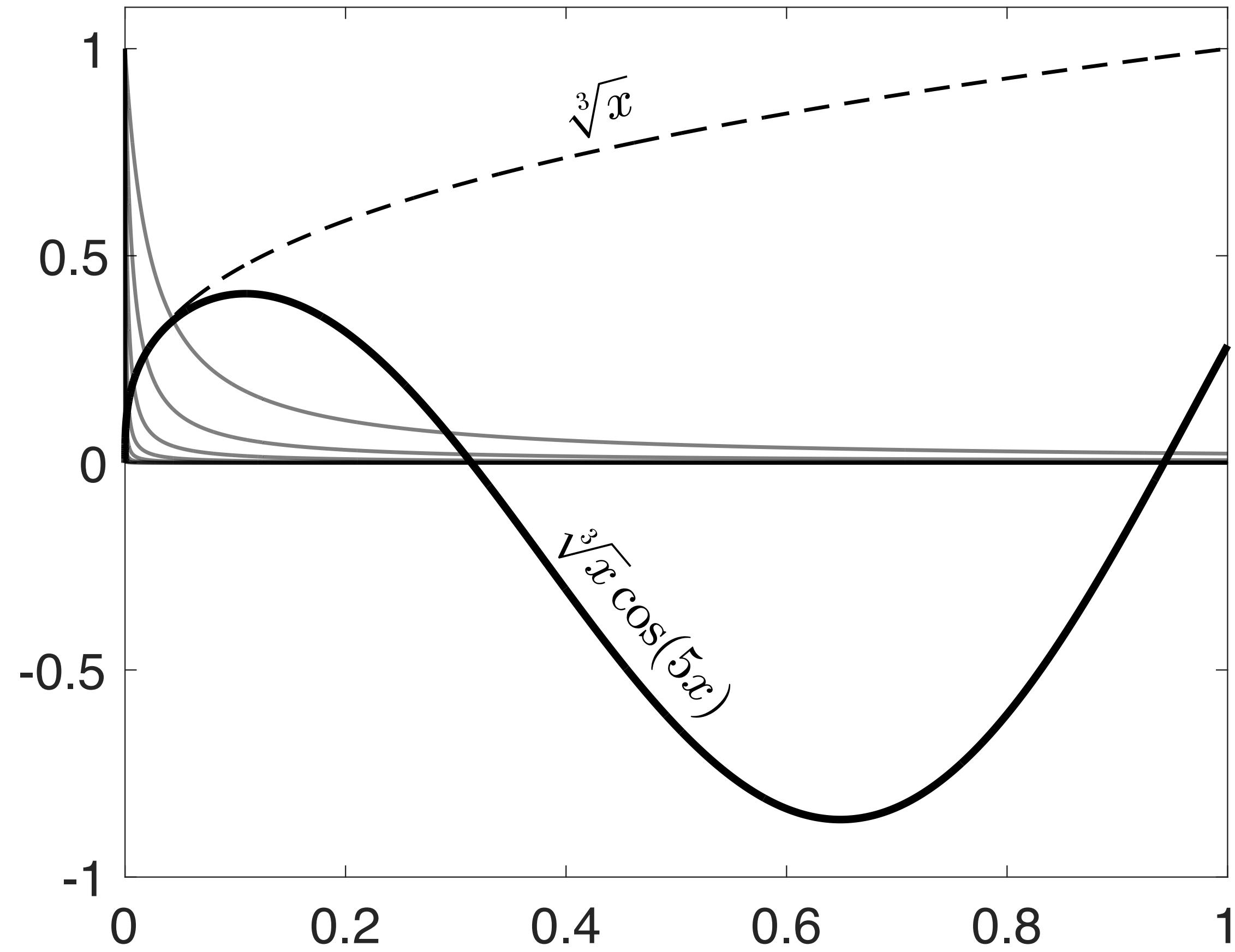
$$\text{rational: } \frac{p_k}{x - p_k}$$

$$s = \log x$$

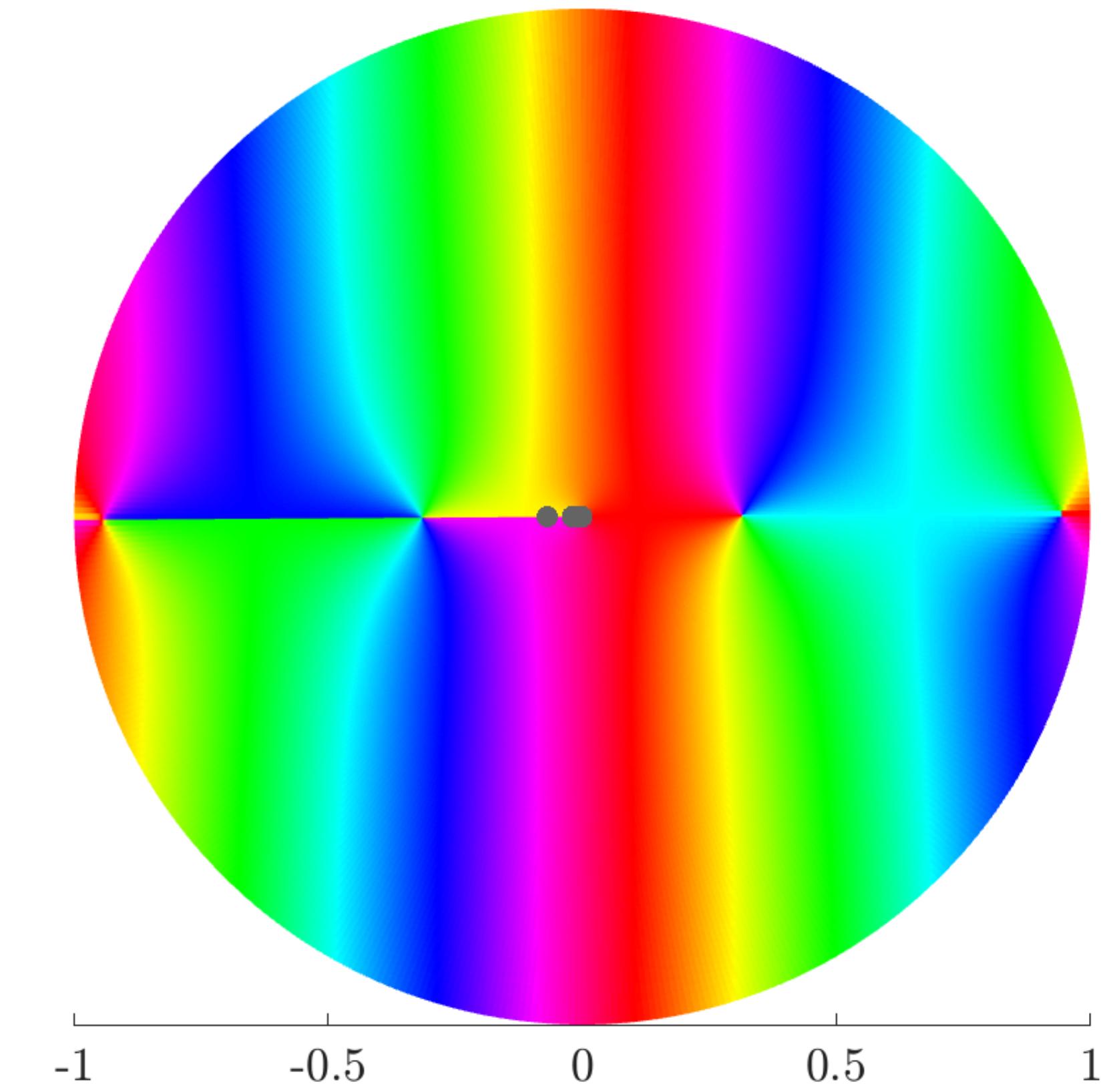
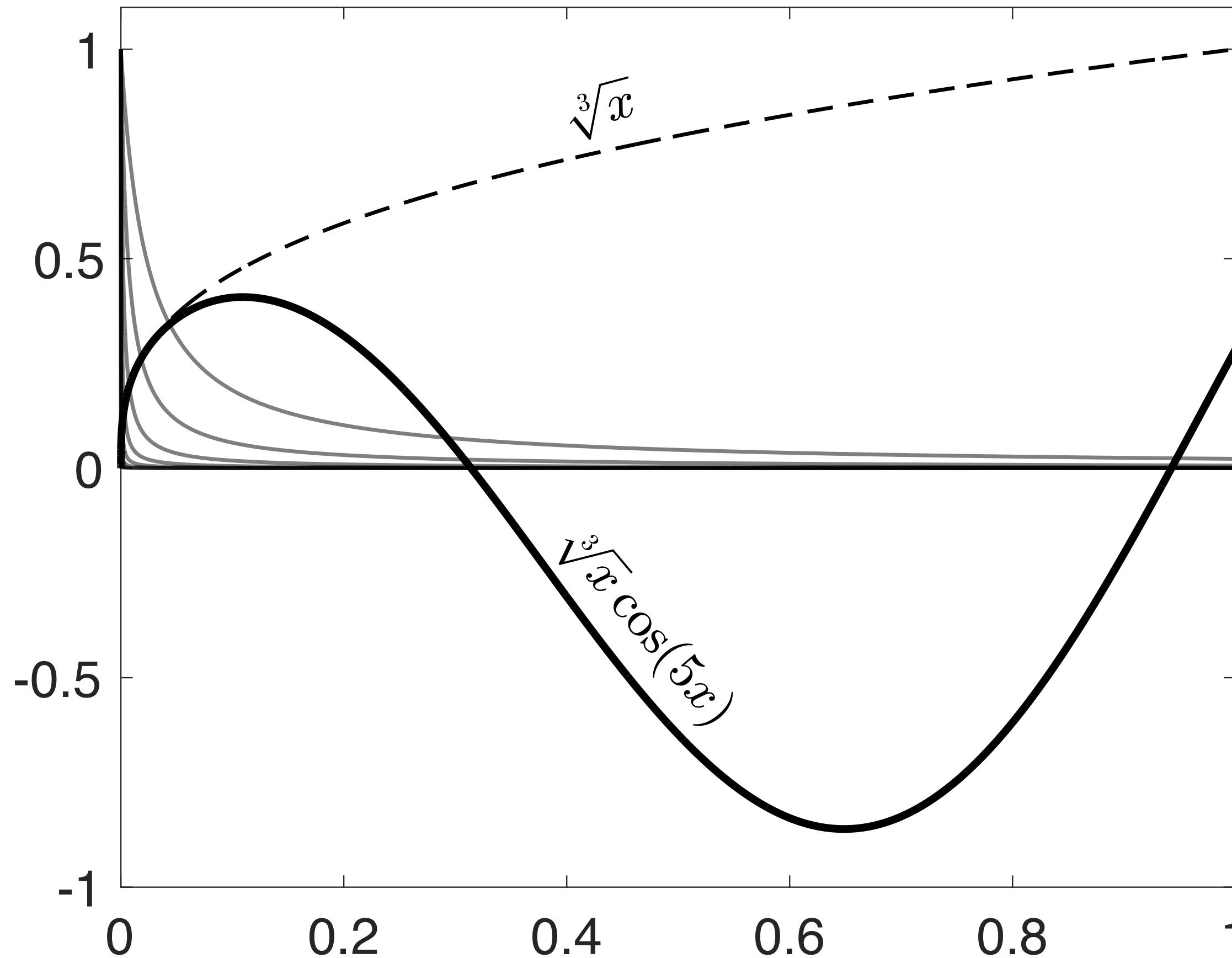
$$\text{sigmoid: } \frac{1}{1 + e^{s-s_k}}$$

(Huybrechs, Trefethen 2024)

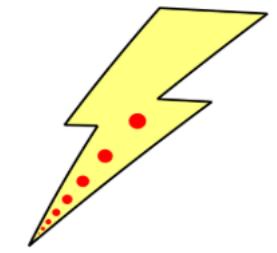




simply use big poles / polynomials to capture the smooth behaviour



Lightning approximation



Given the locations of the singularities $\{z_j\}$ of f ,

$$f(z) \approx r(z) = \underbrace{\sum_{k=1}^{n_1} \frac{a_k p_k}{z - p_k}}_{\text{“LIGHTNING”}} + \underbrace{\sum_{k=0}^{n_2} b_k T_k(z)}_{\text{“POLYNOMIAL”}}$$



$\mathcal{O}(|z - z_j|^{\delta_j})$ as $z \rightarrow z_j$

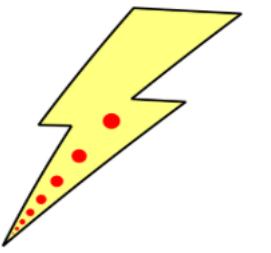


captures singular
behaviour near z_j 's



captures smooth
behaviour

Lightning approximation



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→ finding a_k and b_k can be done via least squares fitting

Root-exponential convergence

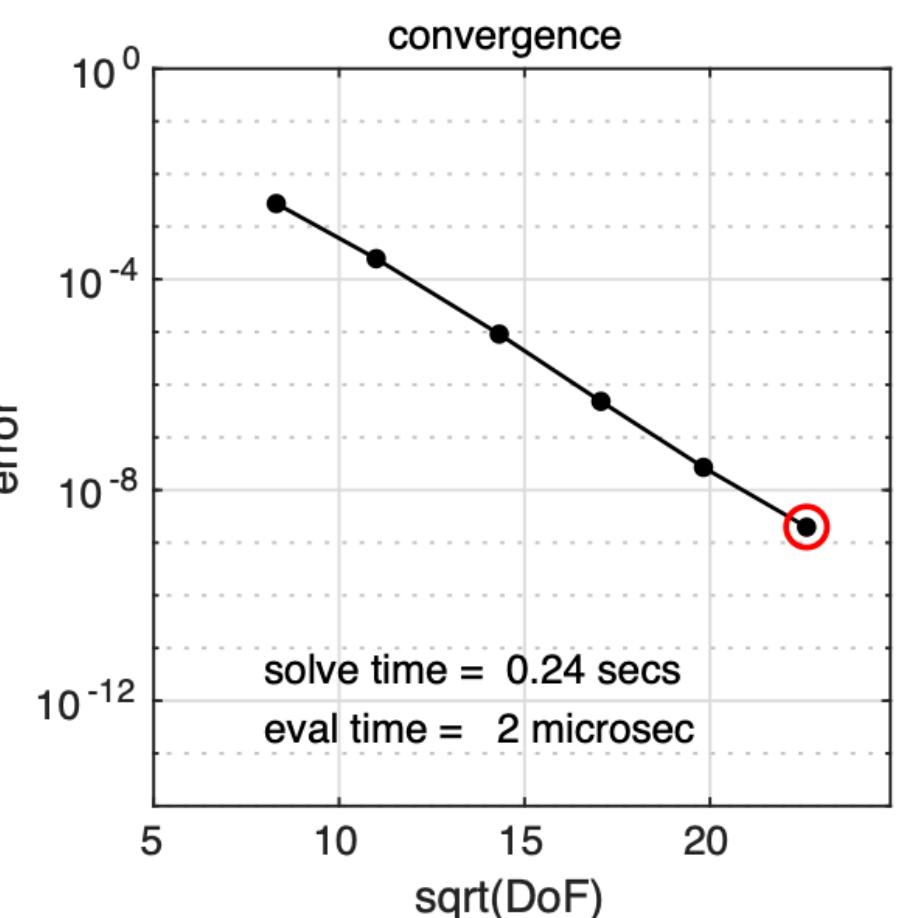
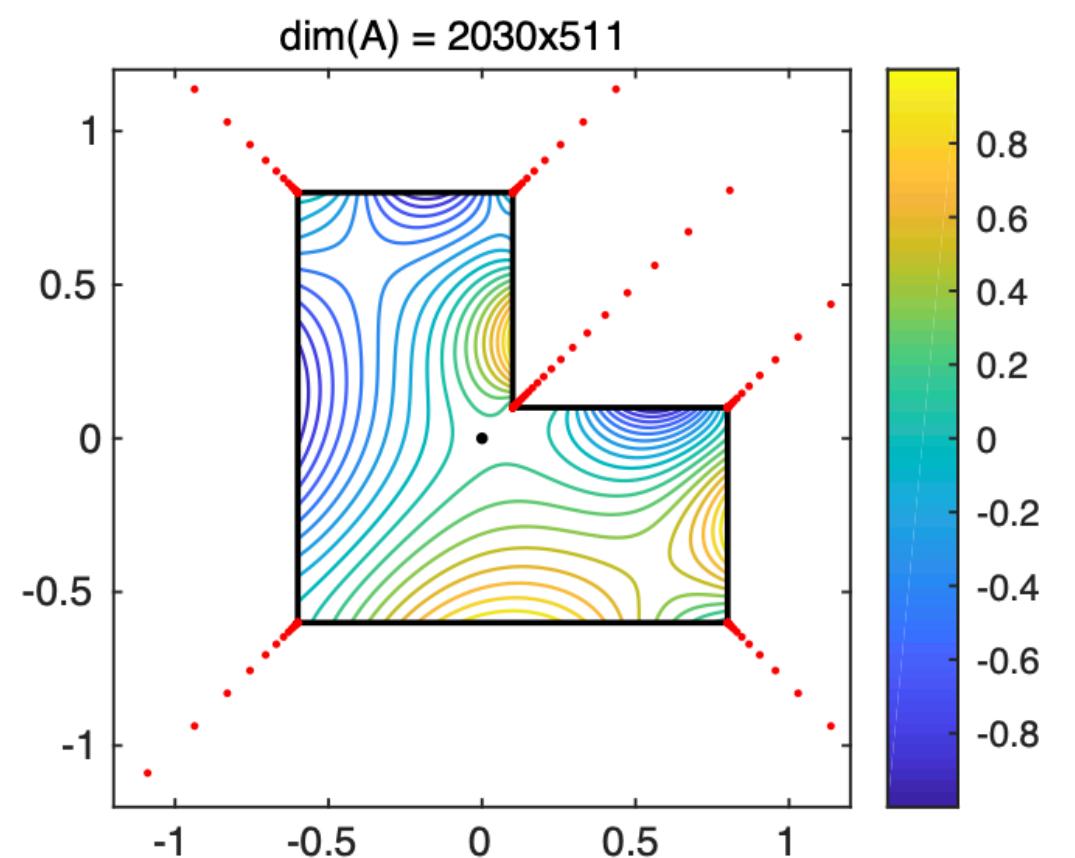
[Gopal and Trefethen, 2019](#)

THEOREM 2.3. Let Ω be a convex polygon with corners w_1, \dots, w_m , and let f be an analytic function in Ω that is analytic on the interior of each side segment and can be analytically continued to a disk near each w_k with a slit along the exterior bisector there. Assume f satisfies $f(z) - f(w_k) = O(|z - w_k|^\delta)$ as $z \rightarrow w_k$ for each k for some $\delta > 0$. There exist degree n rational functions $\{r_n\}$, $1 \leq n < \infty$, such that

$$(2.11) \quad \|f - r_n\|_\Omega = O(e^{-C\sqrt{n}})$$

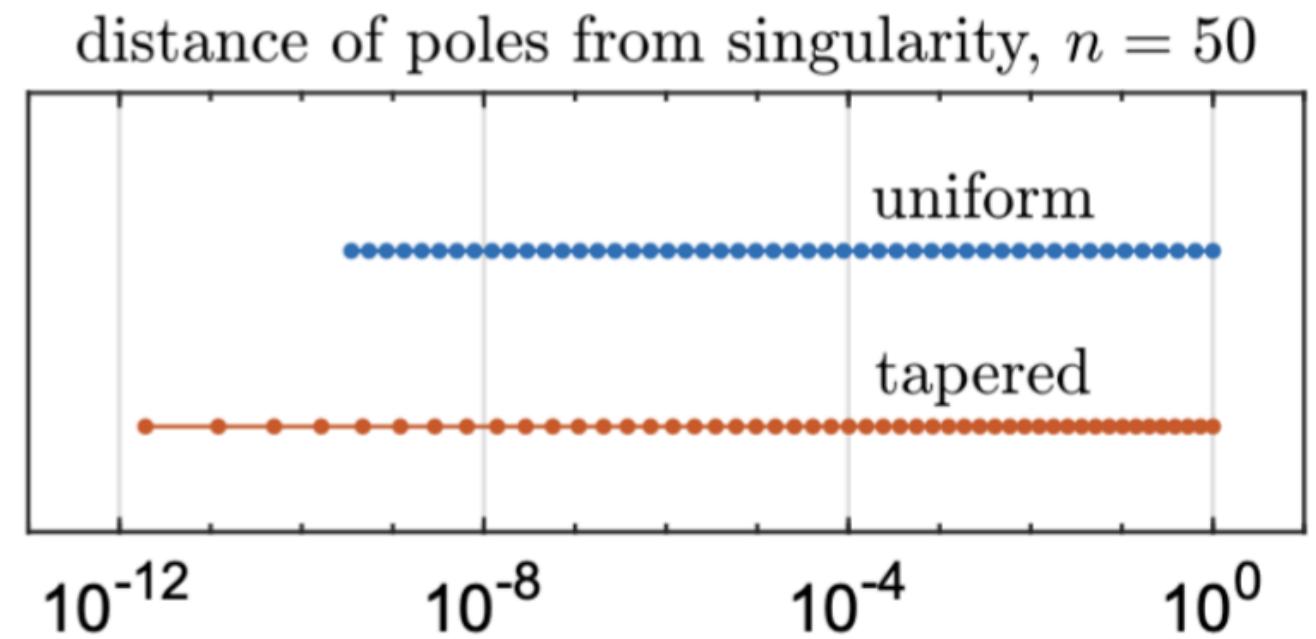
as $n \rightarrow \infty$ for some $C > 0$. Moreover, each r_n can be taken to have finite poles only at points exponentially clustered along the exterior bisectors at the corners, with arbitrary clustering parameter σ as in (2.5), as long as the number of poles near each w_k grows at least in proportion to n as $n \rightarrow \infty$.

$$(2.5) \quad \beta_j = -e^{-\sigma j/\sqrt{n}}, \quad 0 \leq j \leq n - 1$$



Optimisations

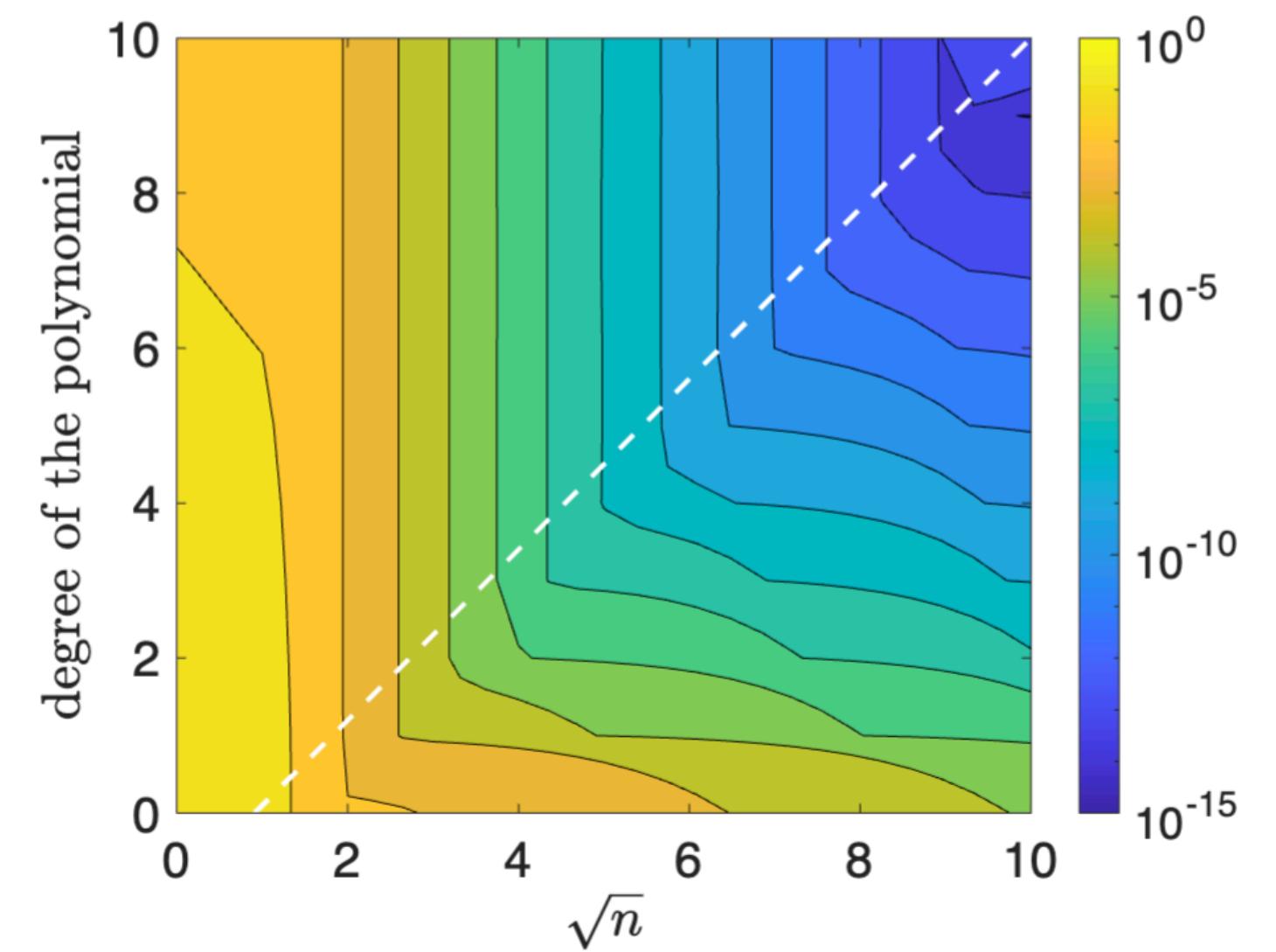
- “tapering” of the poles (Trefethen, Nakatsukasa and Weideman, 2021)



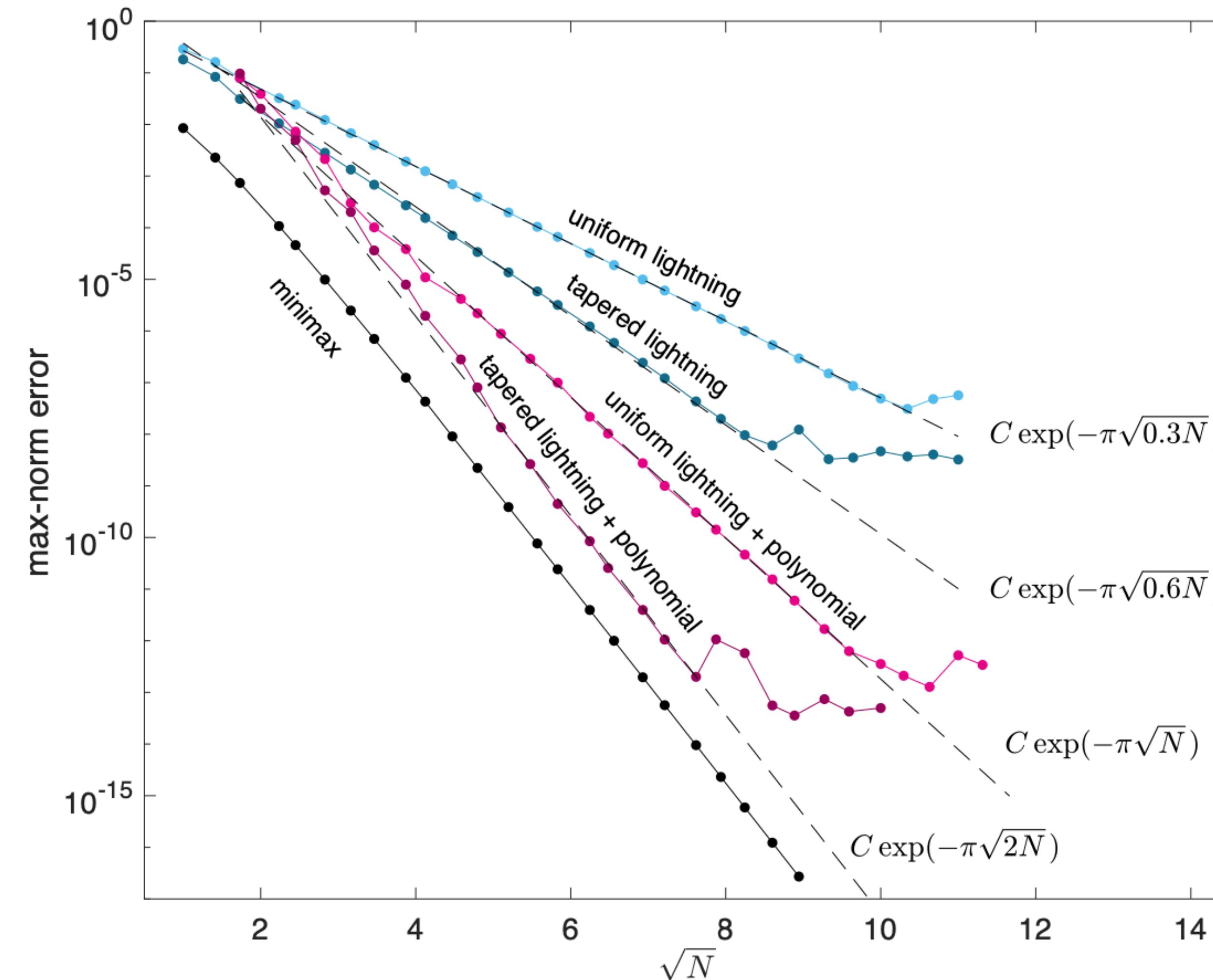
- adding a low-degree polynomial (H., Huybrechs and Trefethen, 2023)

- for singularities of type x^α , use $\sigma = 2\pi/\sqrt{\alpha}$

(H., Huybrechs and Trefethen, 2023) + (Xiang, Yang and Wu, 2024)



Optimal convergence rate



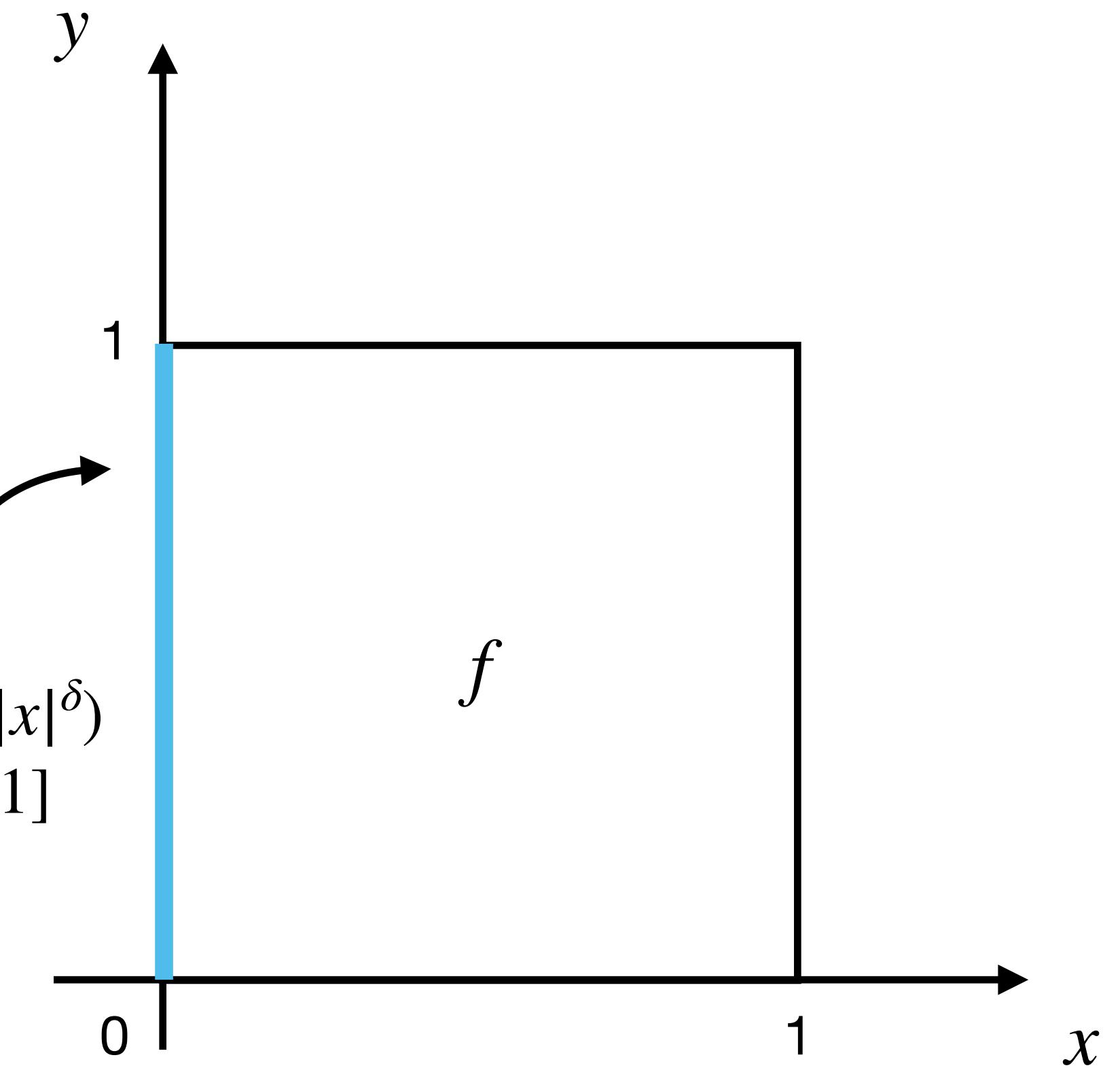
(H., Huybrechs and Trefethen, 2023)

Multivariate case

→ exploring multivariate lightning approximations (Boullé, H. and Huybrechs, 2024)

$$f(x, y) \approx \sum_k \frac{a_k(y)p_k}{x - p_k} + b(x, y)$$

$\exists \delta > 0 : f(x, y) = \mathcal{O}(|x|^\delta)$
as $x \rightarrow 0, \forall y \in [0, 1]$



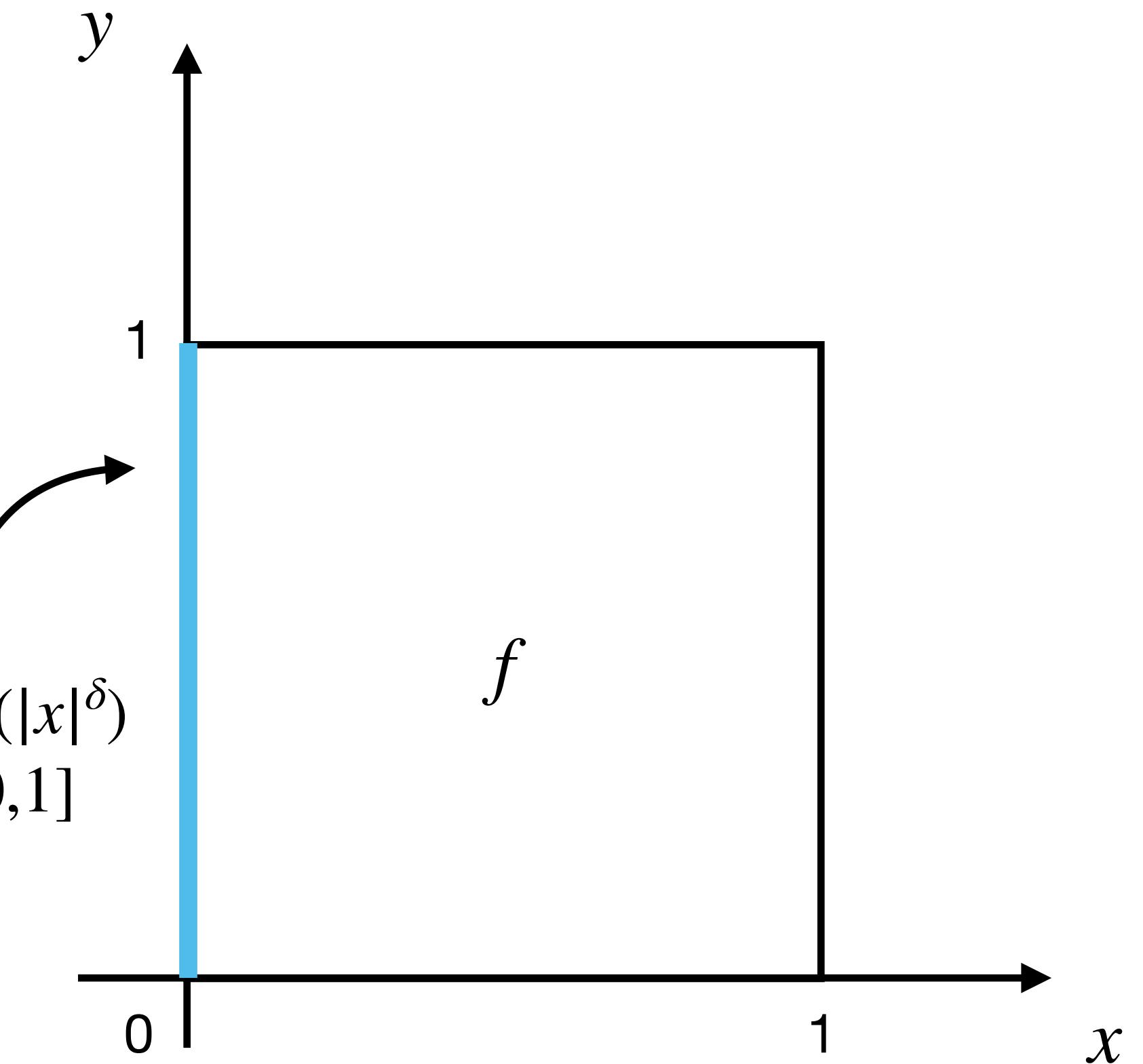
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Remember: poles could be chosen independently of the type of singularity

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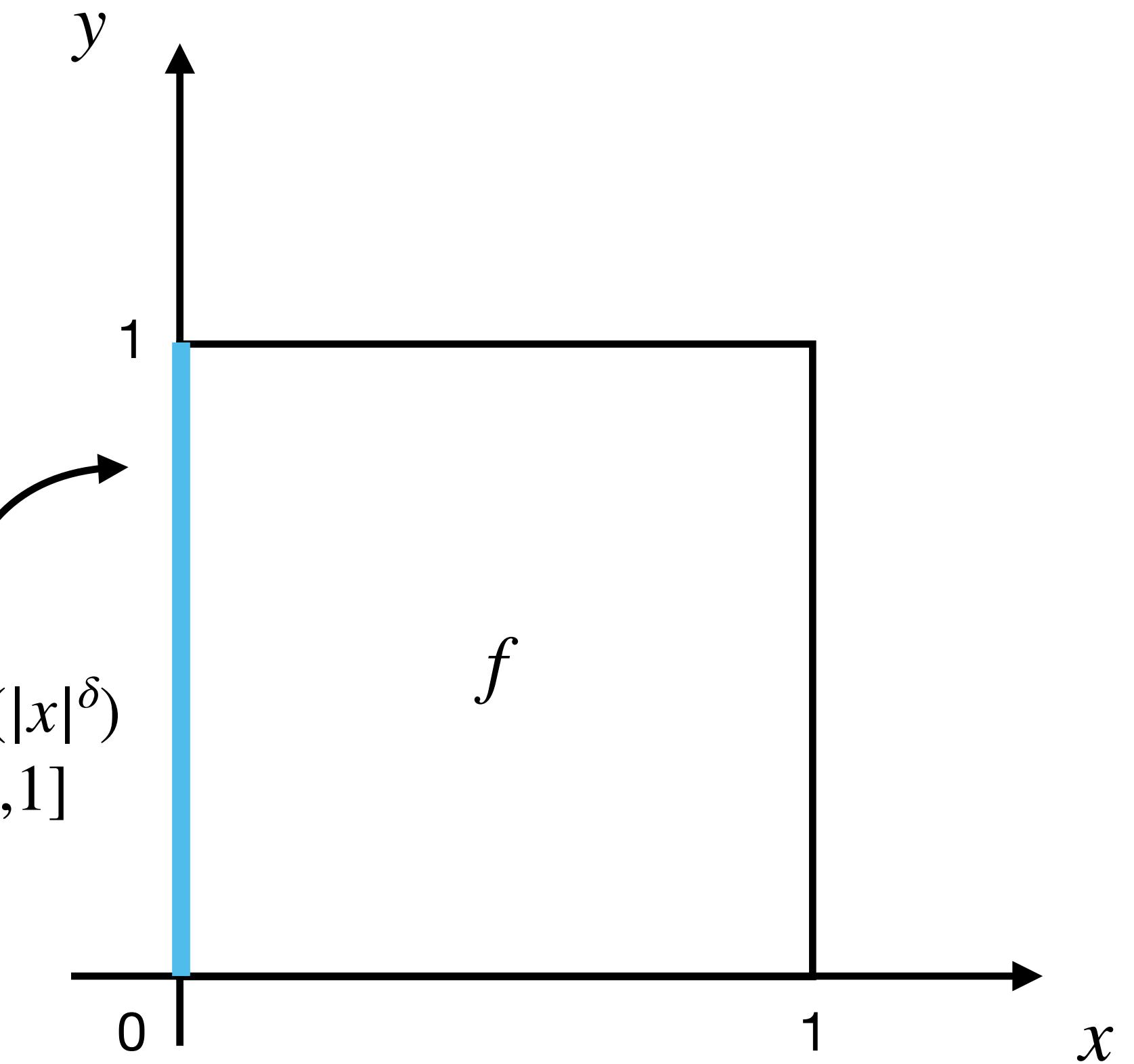
Variation along curve of singularities

→ low degree polynomial

$$f(x, y) \approx \sum_k \frac{a_k(y)p_k}{x - p_k} + b(x, y)$$

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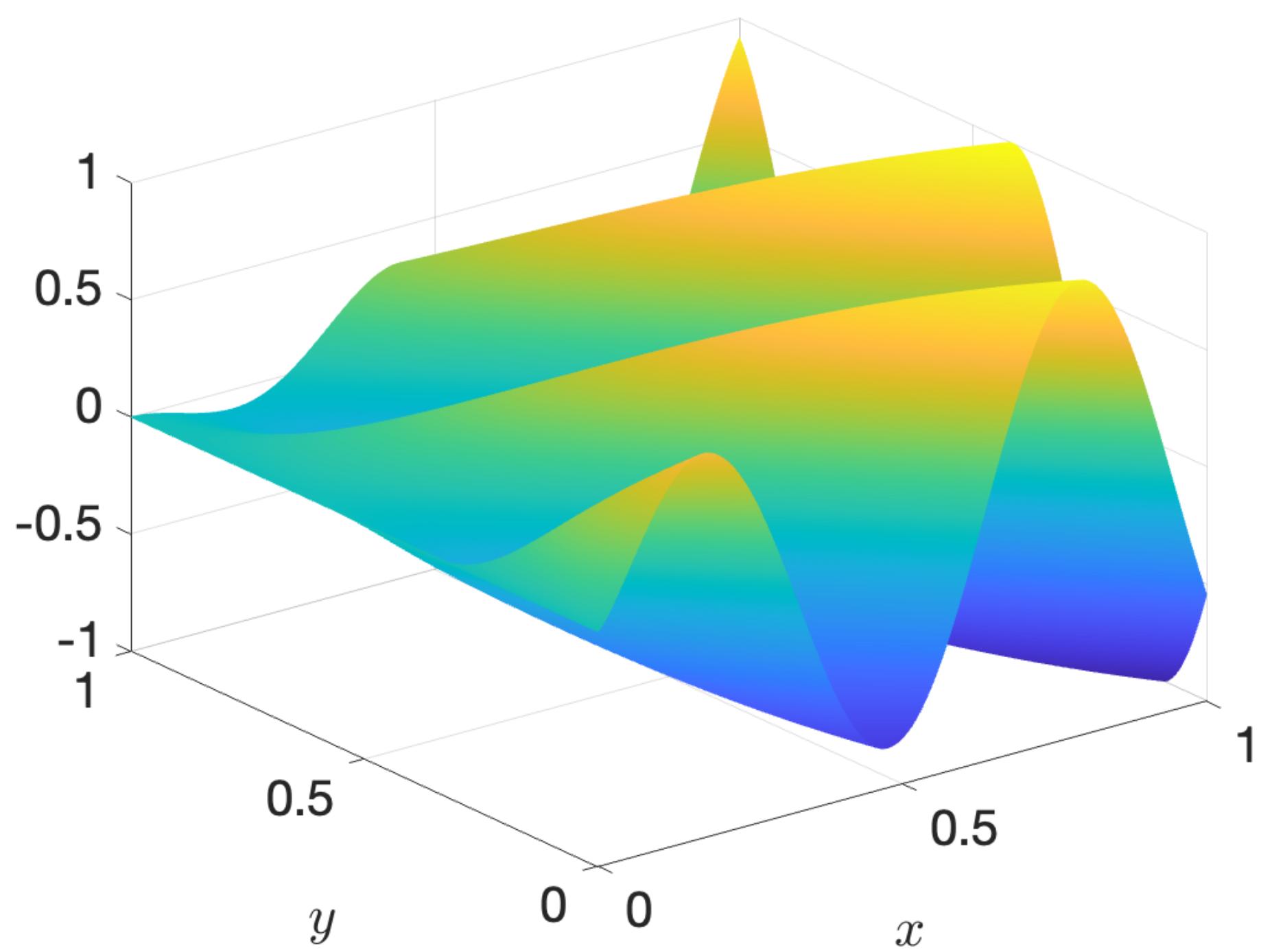
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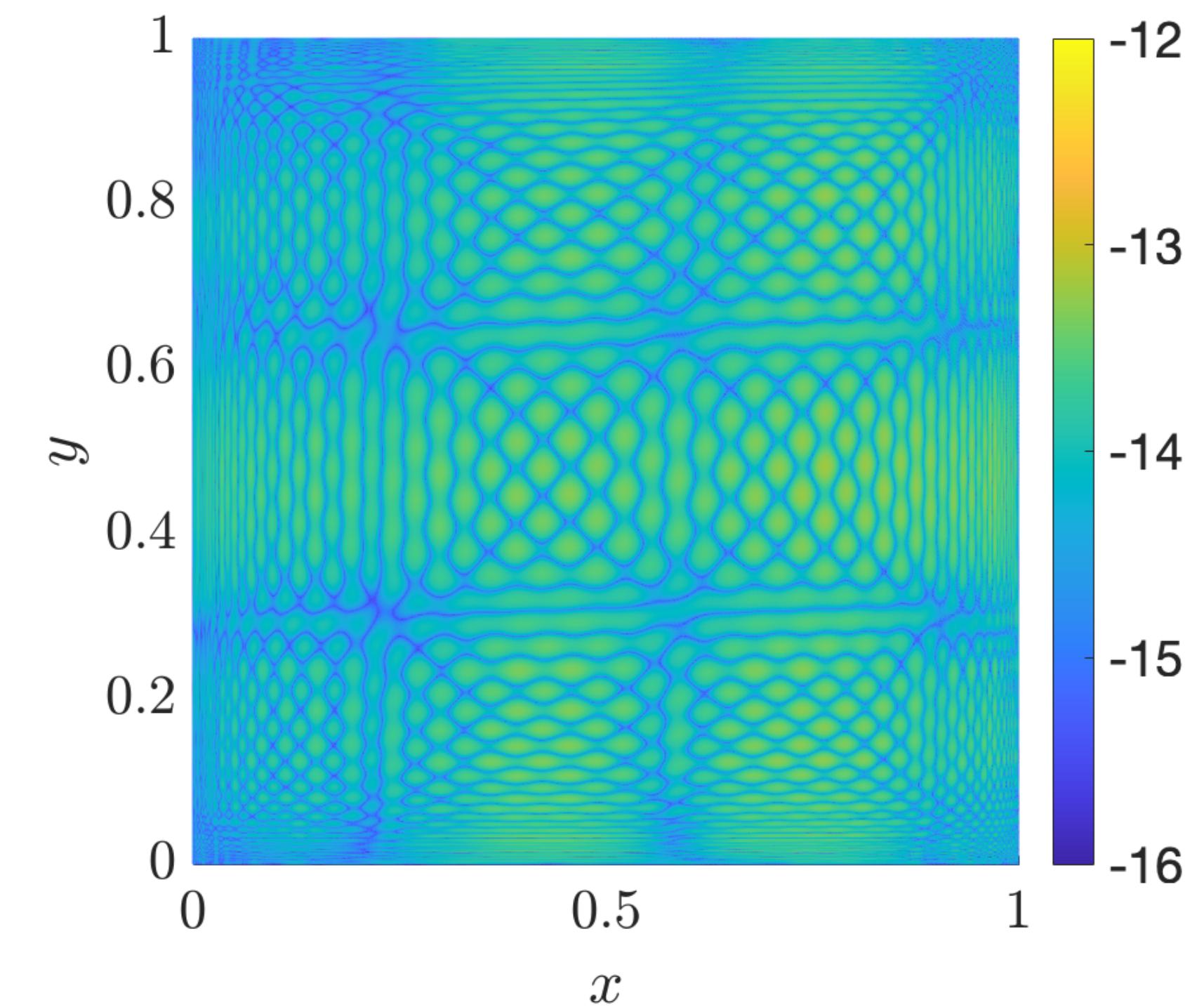
Multivariate case

$$f(x, y) \approx \sum_k \frac{a_k(y)p_k}{x - p_k} + b(x, y)$$

$$f = x^{1/4+y} \sin(10(x + y))$$



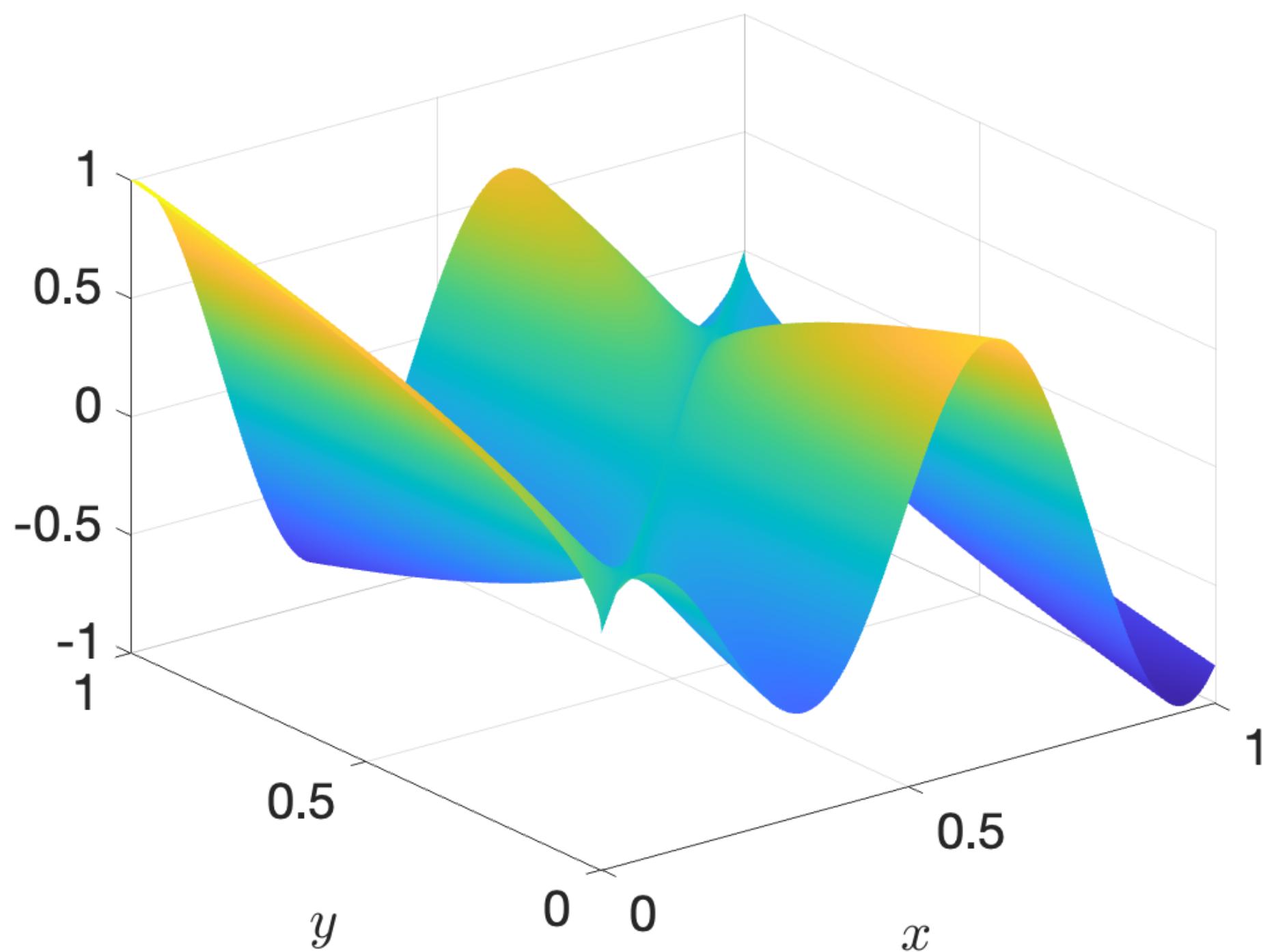
$$\log_{10}(\text{error})$$



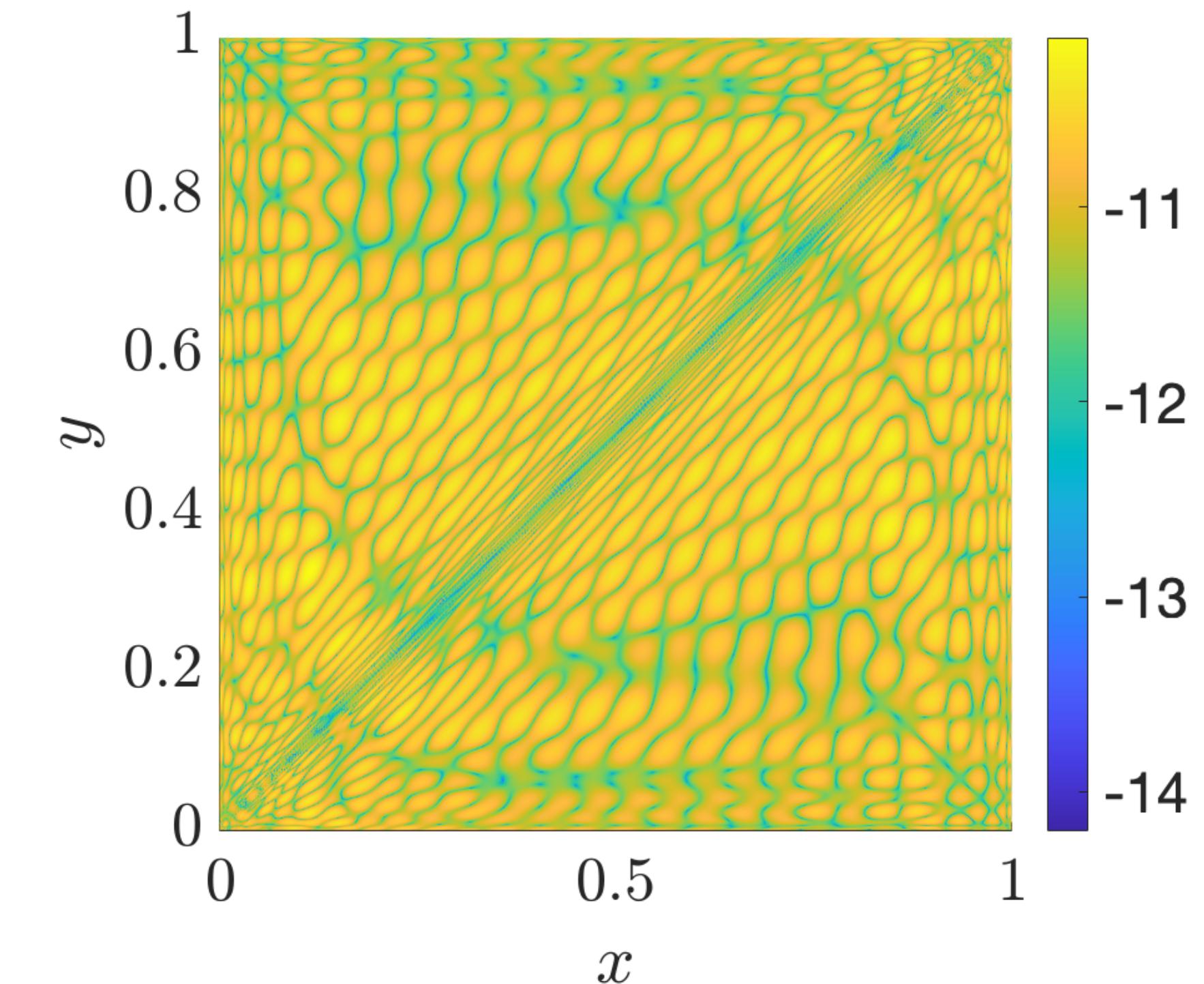
Multivariate case

$$f(x, y) \approx \sum_k \frac{a_k(x + y)p_k}{(y - x) \pm ip_k} + b(x, y)$$

$$f = \sqrt{|x - y|} \cos(10x)$$



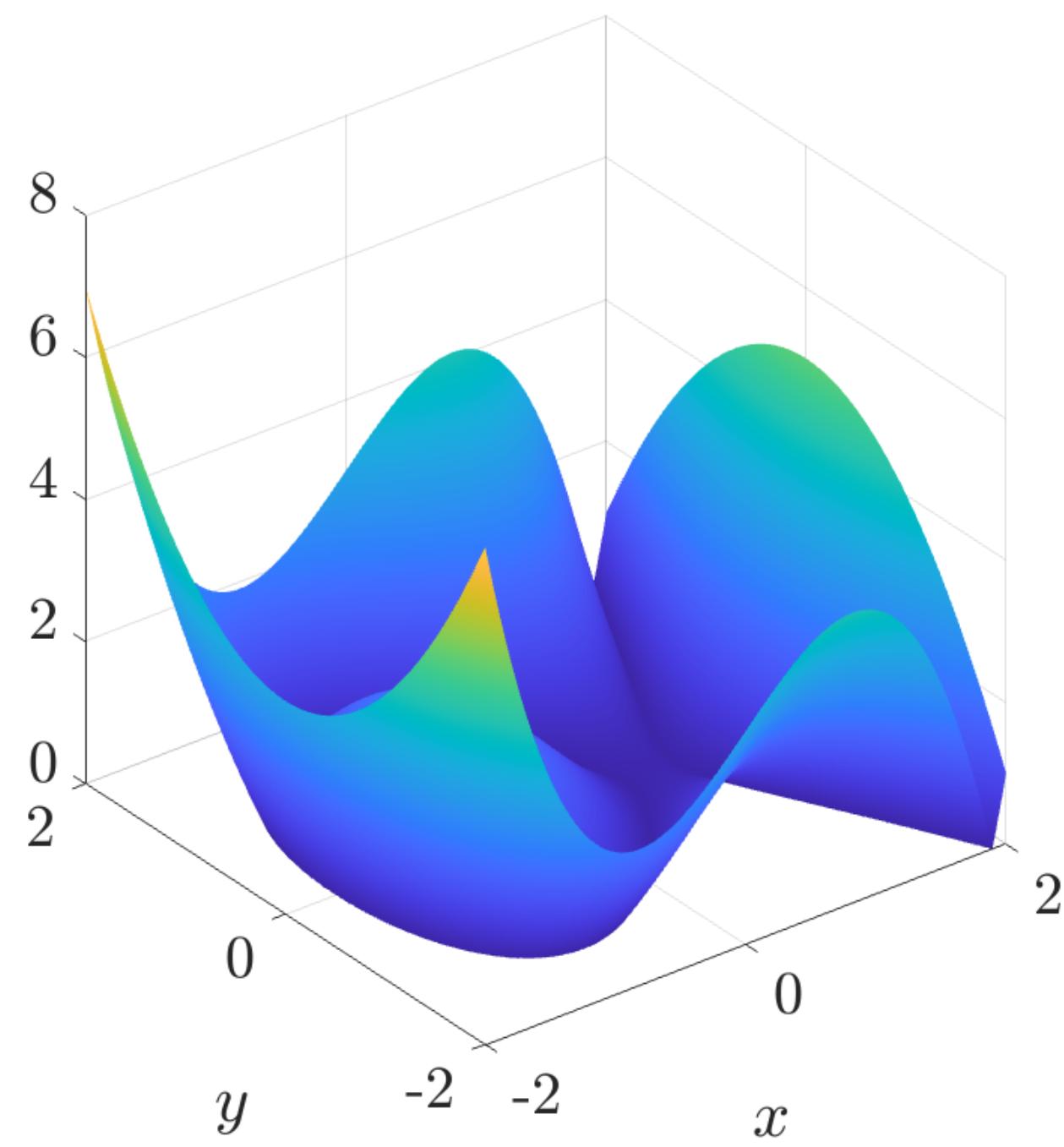
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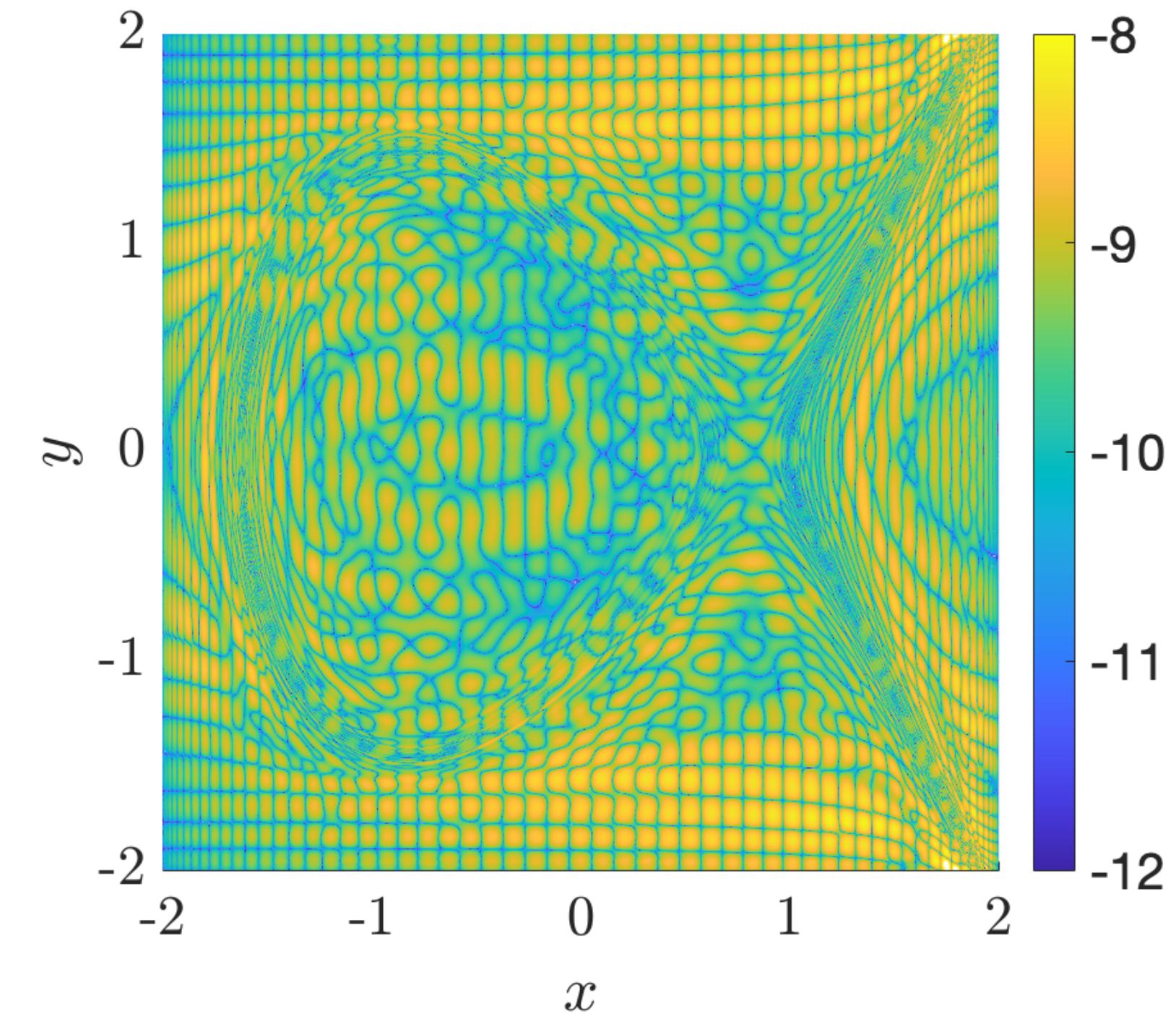
Multivariate case

$$f(x, y) \approx \sum_k \frac{a_k(x, y)p_k}{C(x, y) \pm ip_k} + b(x, y)$$

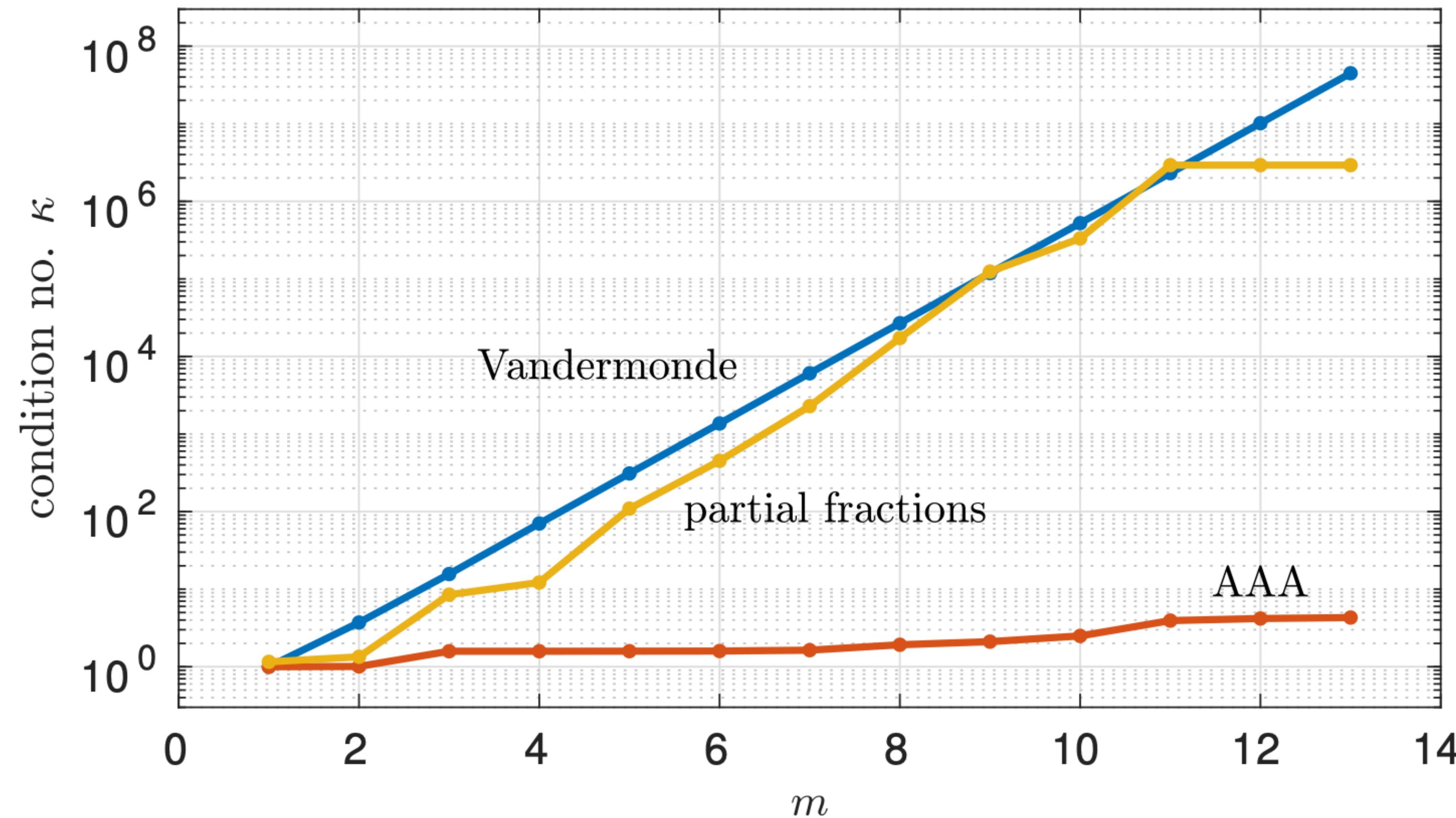
$$f = |x^3 - 2x + 1 - y^2| = |C(x, y)|$$



$\log_{10}(\text{error})$

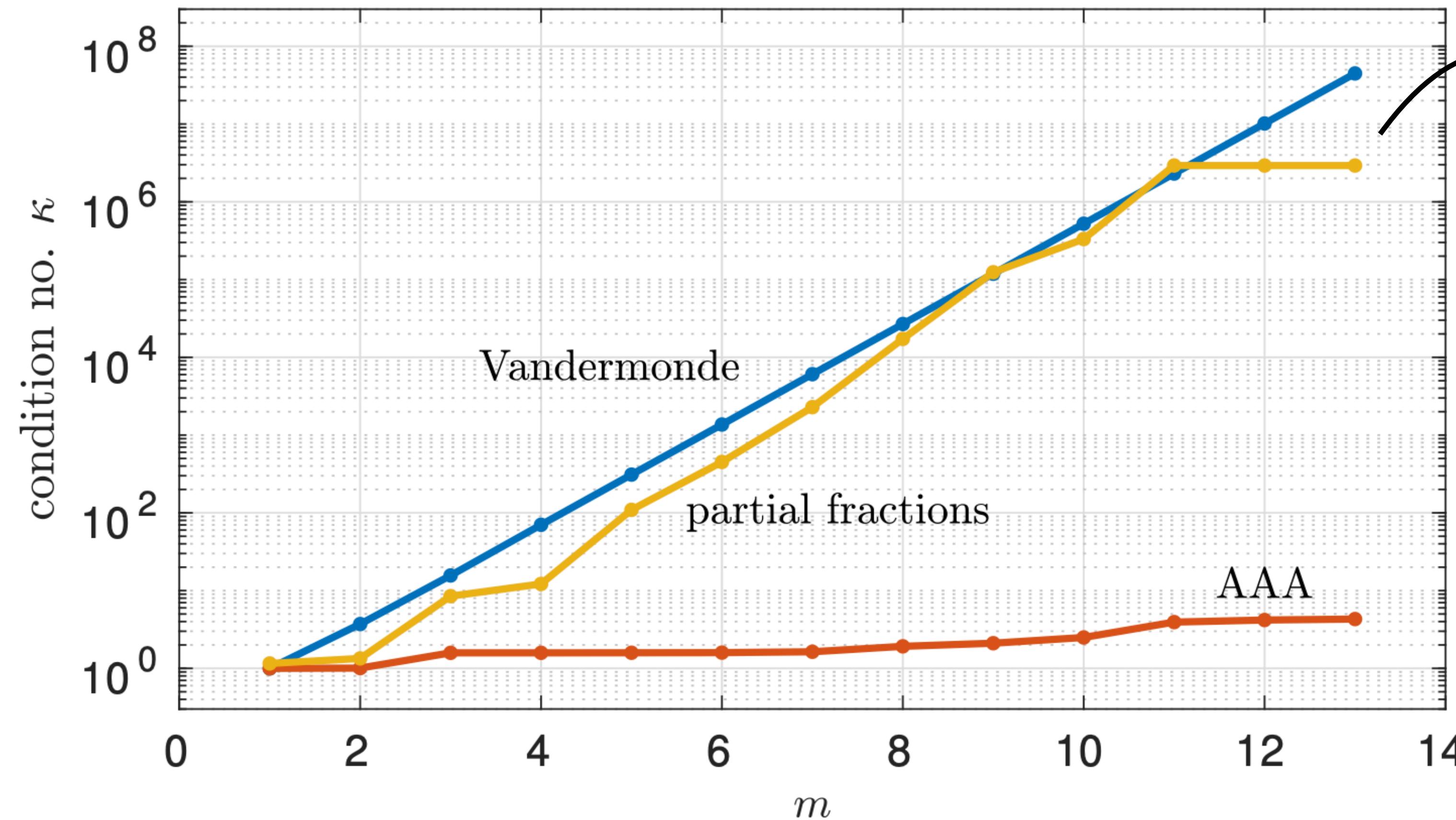


On the partial fractions representation



(Nakatsukasa, Sète and Trefethen, 2018)

On the partial fractions representation



This means that the basis
is close to being linearly
dependent.

Is this a problem?

Sensitivity to data

Given sampled data $(b)_i = f(x_i)$, we compute $\hat{f} = \sum_k \hat{c}_k \phi_k$ where

$$\hat{c} = \arg \min_c \|Ac - b\|_2^2 \quad \text{with } (A)_{i,j} = \phi_j(x_i)$$

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The (absolute) condition number of $b \mapsto \hat{f}$ satisfies

$$\kappa \leq \frac{1}{C} \quad \text{where } C\|\nu\| \leq \|\{\nu(x_i)\}_i\|_2, \quad \forall \nu \in \text{span}(\phi_k)$$

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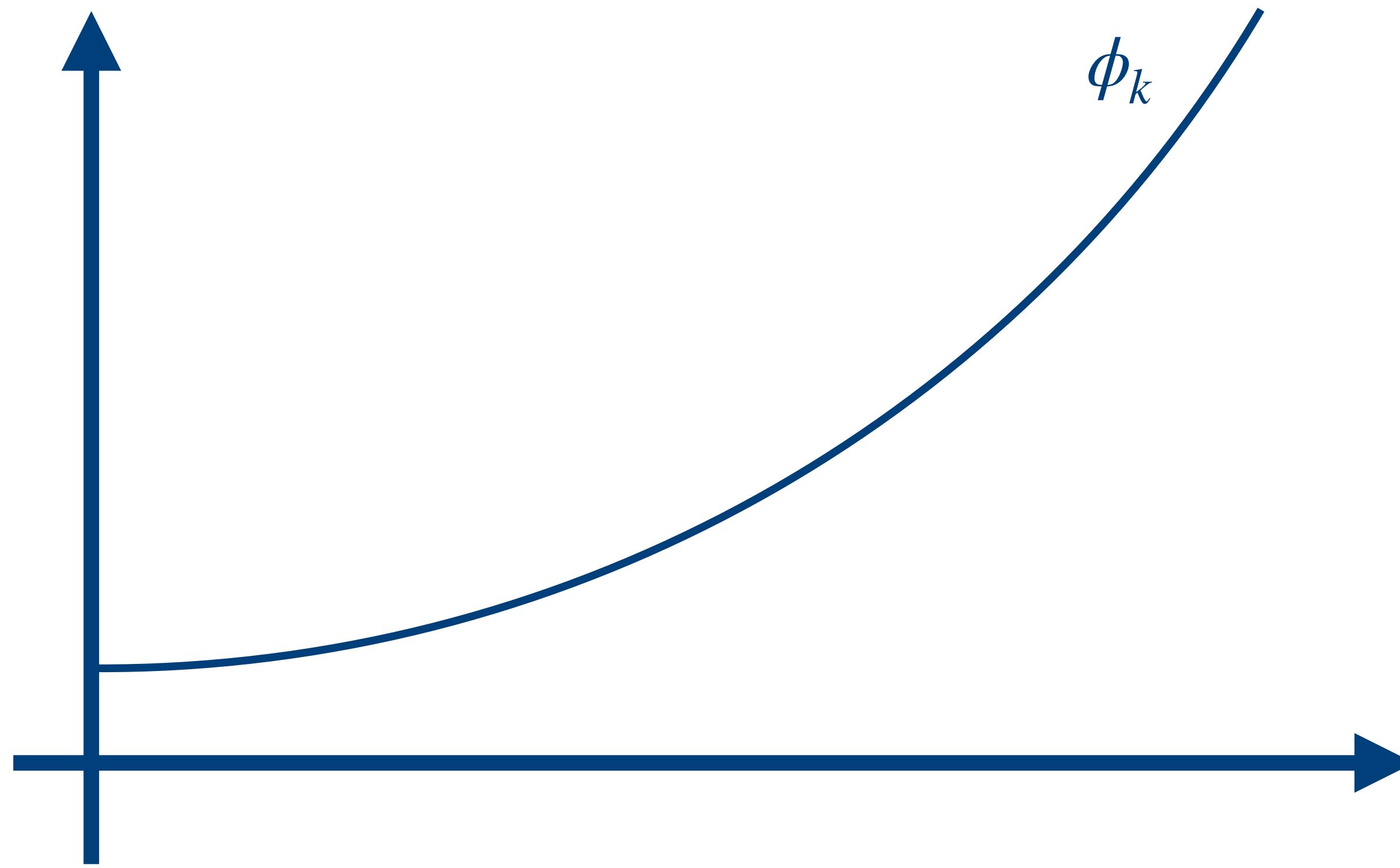
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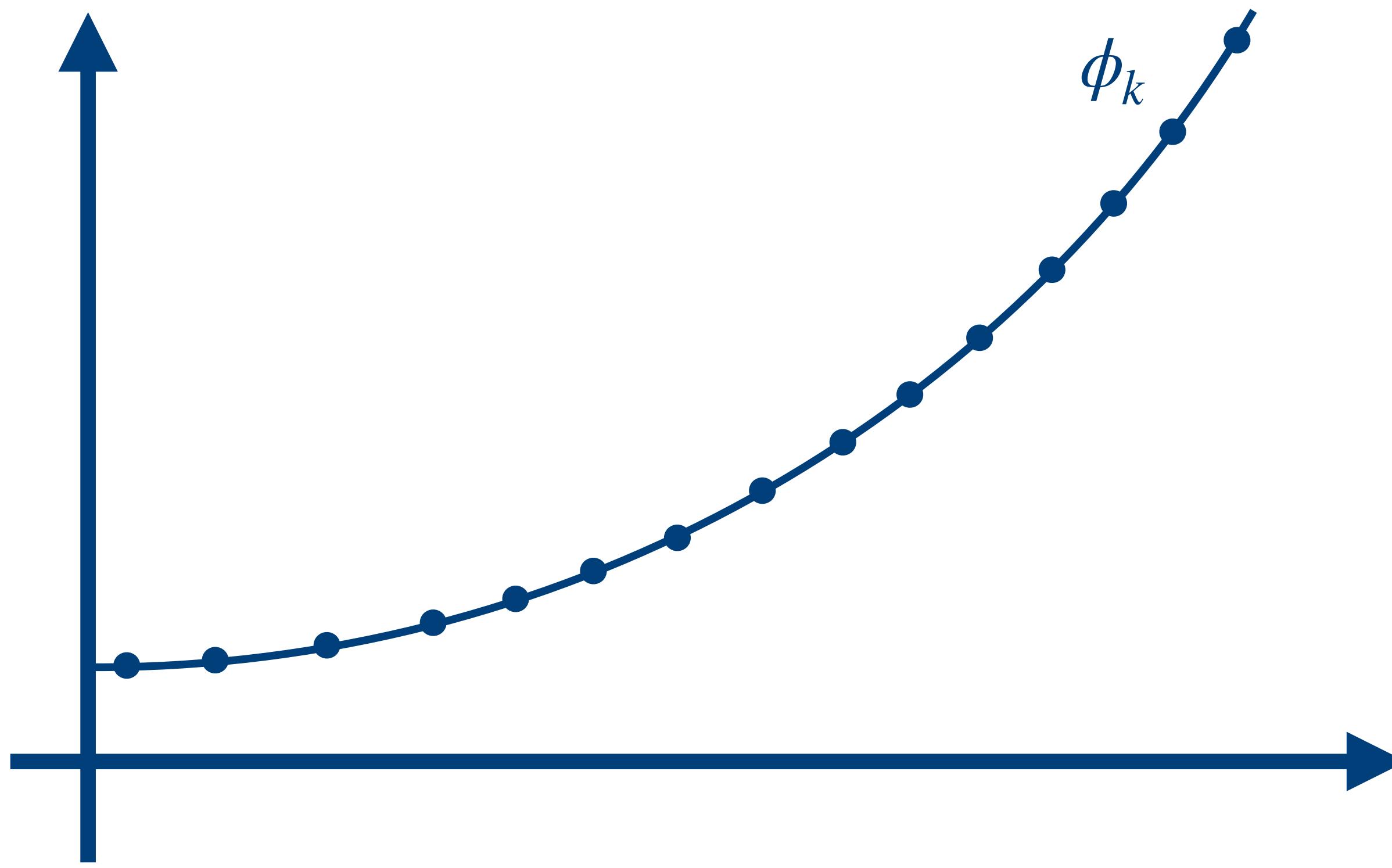
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This does NOT depend on (the condition number of) the basis ϕ_k

Numerical approximation space

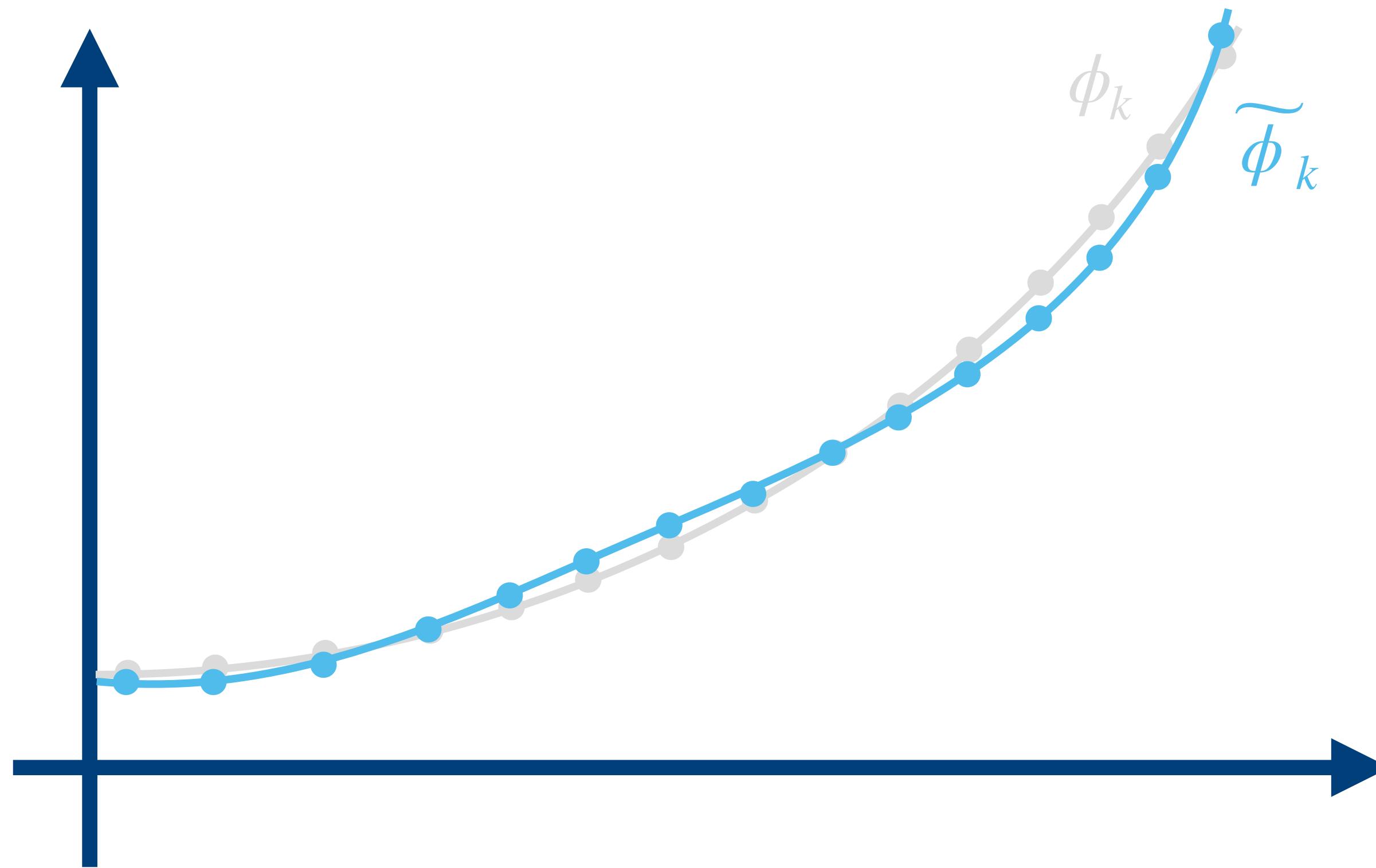


Numerical approximation space



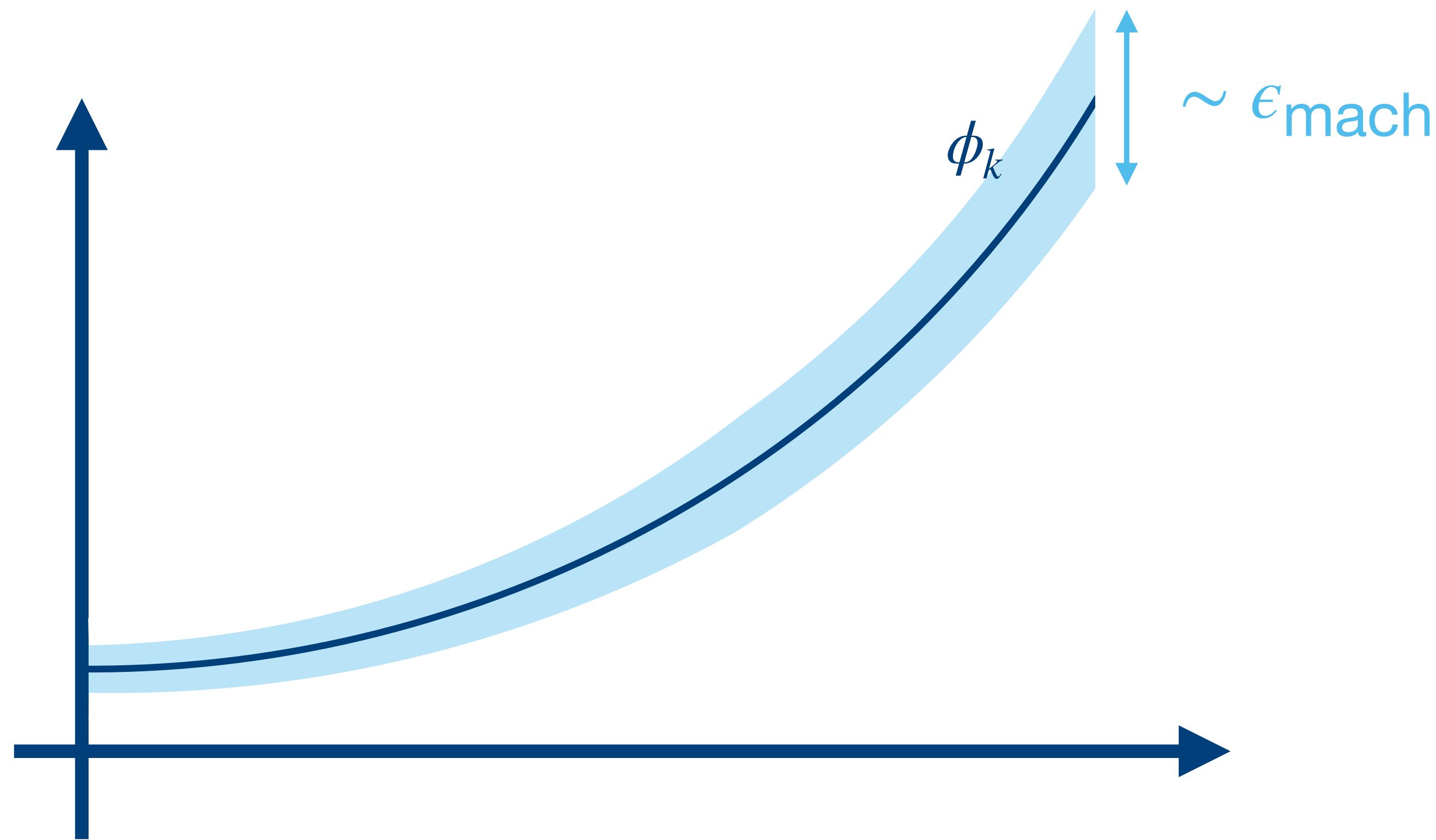
Numerical approximation space

on a computer we work with $\tilde{\phi}_k$ instead of ϕ_k



Numerical approximation space

we don't know $\tilde{\phi}_k$ but we know it lies close to ϕ_k



Numerical approximation space

best approximation error in $\text{span}(\phi_k)$

$$\inf_c \left(\|f - \sum_k c_k \phi_k\| \right)$$

best approximation error in $\text{span}(\tilde{\phi}_k)$

$$\leq \inf_c \left(\|f - \sum_k c_k \phi_k\| + \epsilon_{\text{mach}} B \|c\|_2 \right)$$

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$$\sup_{\|c\|_2=1} \|\sum_k c_k \phi_k\|$$

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- in practice: use ℓ^2 -regularization, mind the scaling of ϕ_k

Numerical approximation space

(Adcock and Huybrechs, 2019/2020), (H. and Huybrechs, 2025)

best approximation error in $\text{span}(\phi_k)$

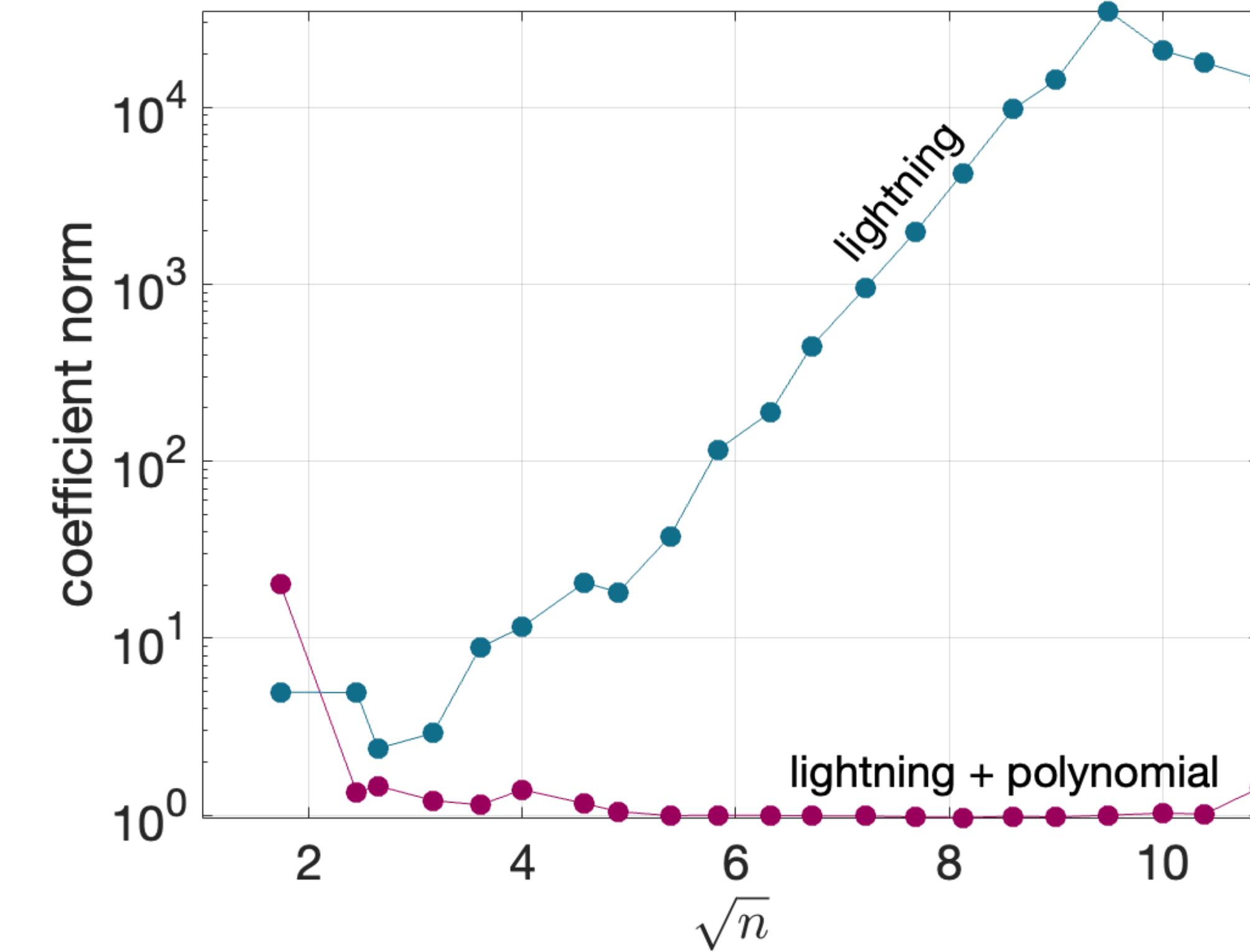
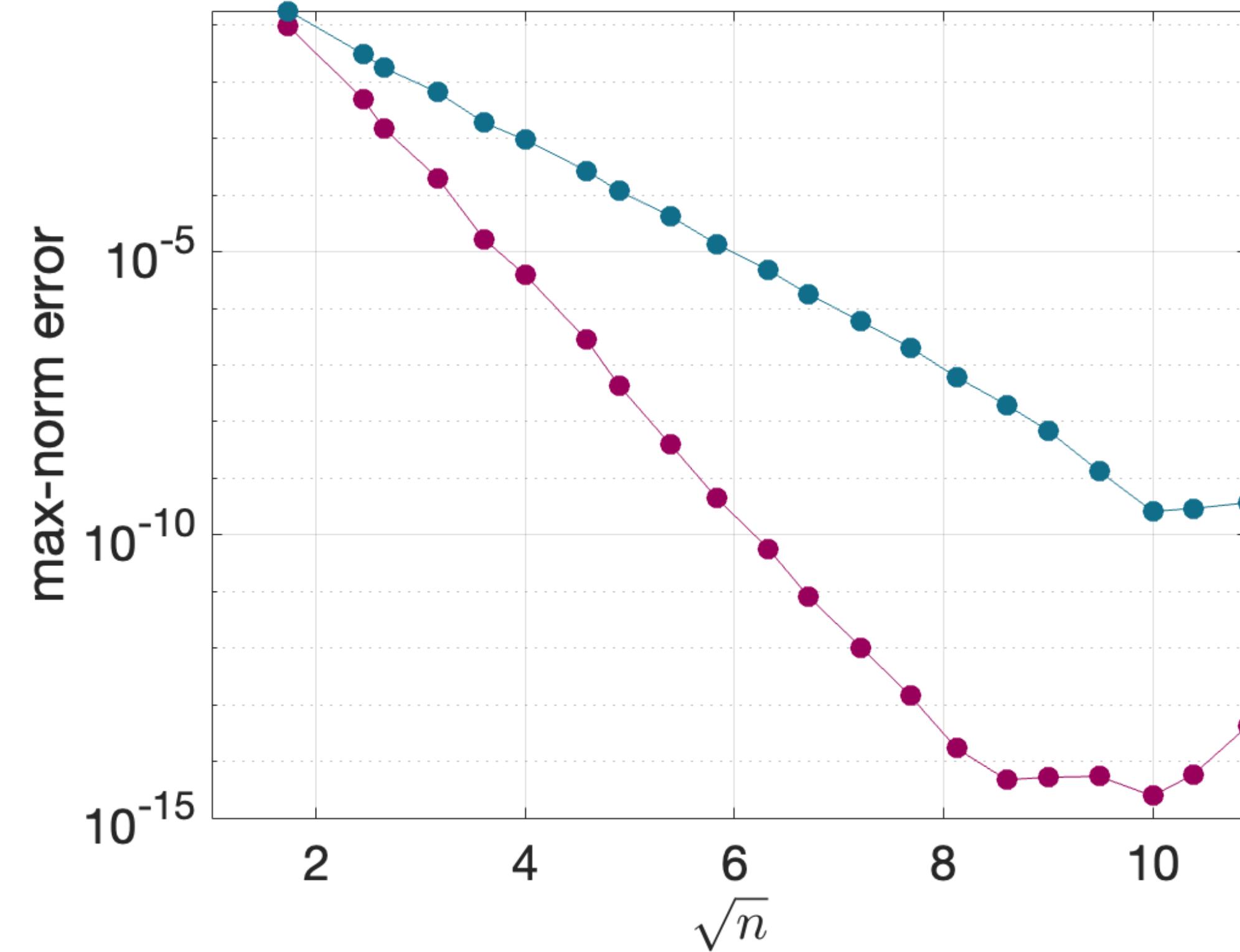
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best approximation error in $\text{span}(\tilde{\phi}_k)$

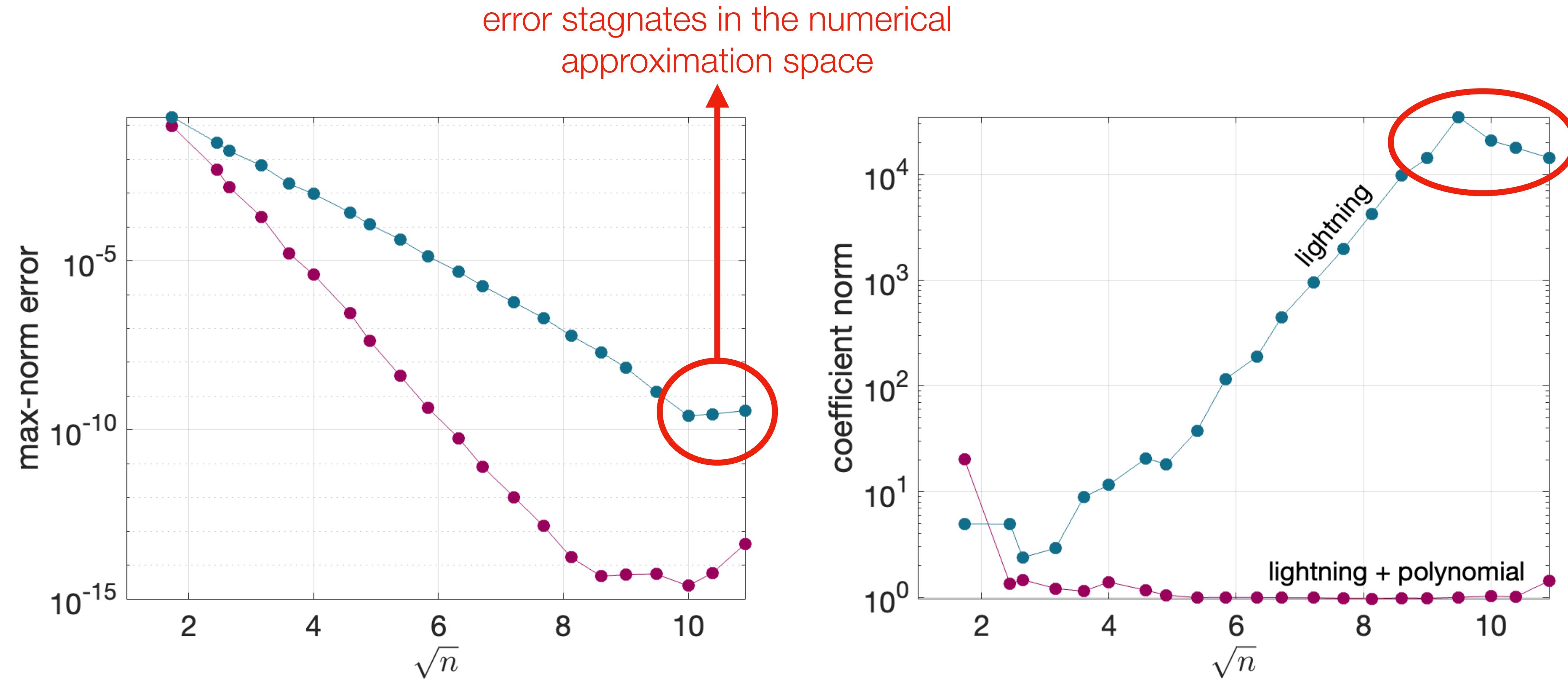
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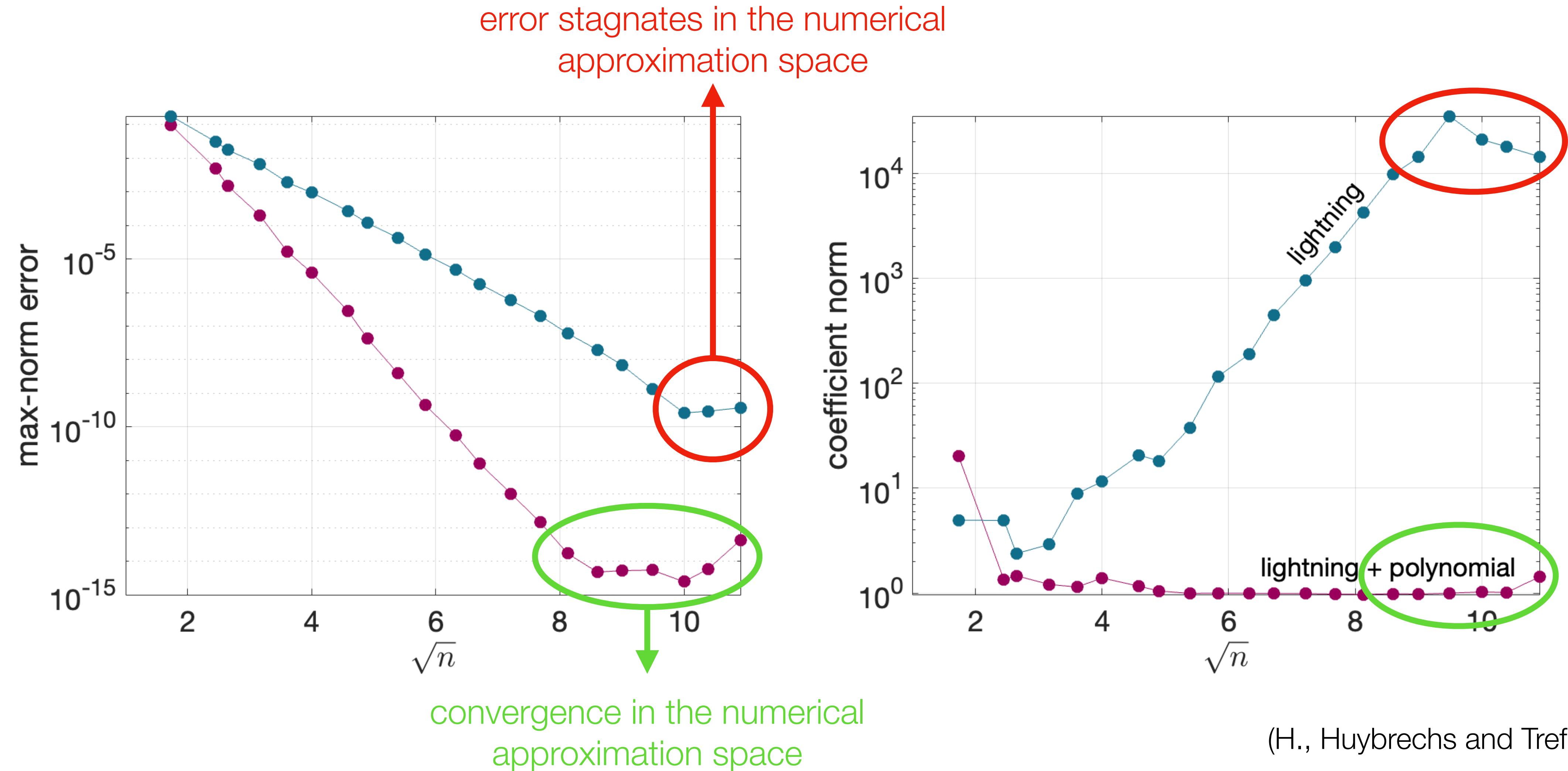


Numerical approximation space



(H., Huybrechs and Trefethen, 2023)

Numerical approximation space



Conclusions

Given the locations of the singularities of a function, we can construct root-exponentially converging rational approximations via least squares fitting

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Given the locations of the singularities of a function, we can construct root-exponentially converging rational approximations via least squares fitting

- many mysteries have been solved in 1D
- lots of exploring to do in higher dimensions
- the lightning basis is “ill-conditioned” yet accurate approximations exist in the numerical approximation space → no need to panic ☺