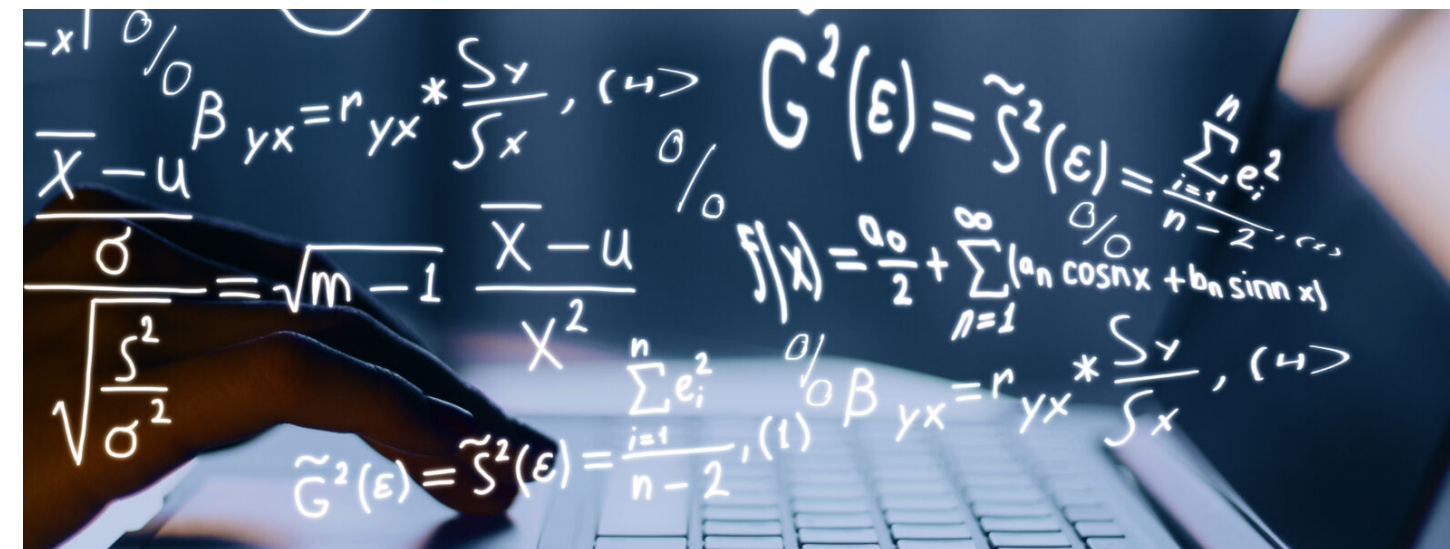


Lattice QCD sign problem as an inverse problem: analytic continuation, Lee-Yang zeros and all that

Francesco Di Renzo (University of Parma and INFN)

Analytic structure of QCD and Yang-Lee edge singularity

ECT*, Trento, 11/09/2025



In collaboration with P. Dimopoulos, M. Aliberti and D. Gavriel (Parma)
(... and Bielefeld Parma Collaboration ...)

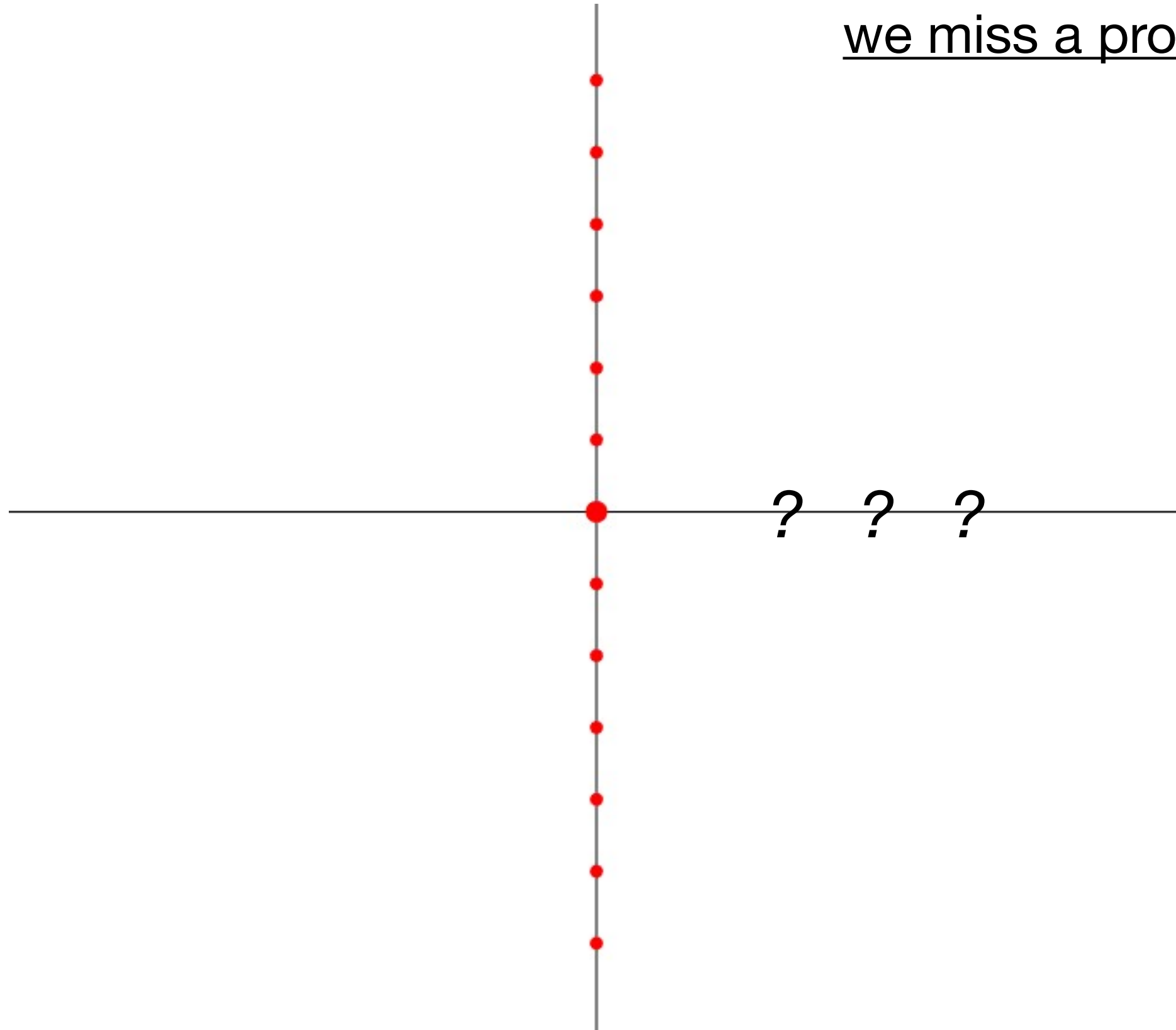


UNIVERSITÀ
DI PARMA



SIGN PROBLEM for finite density Lattice QCD:

we miss a properly defined (positive) measure in the path integral! ... no MC simulation
(... *but everything is fine on the imaginary axis*)

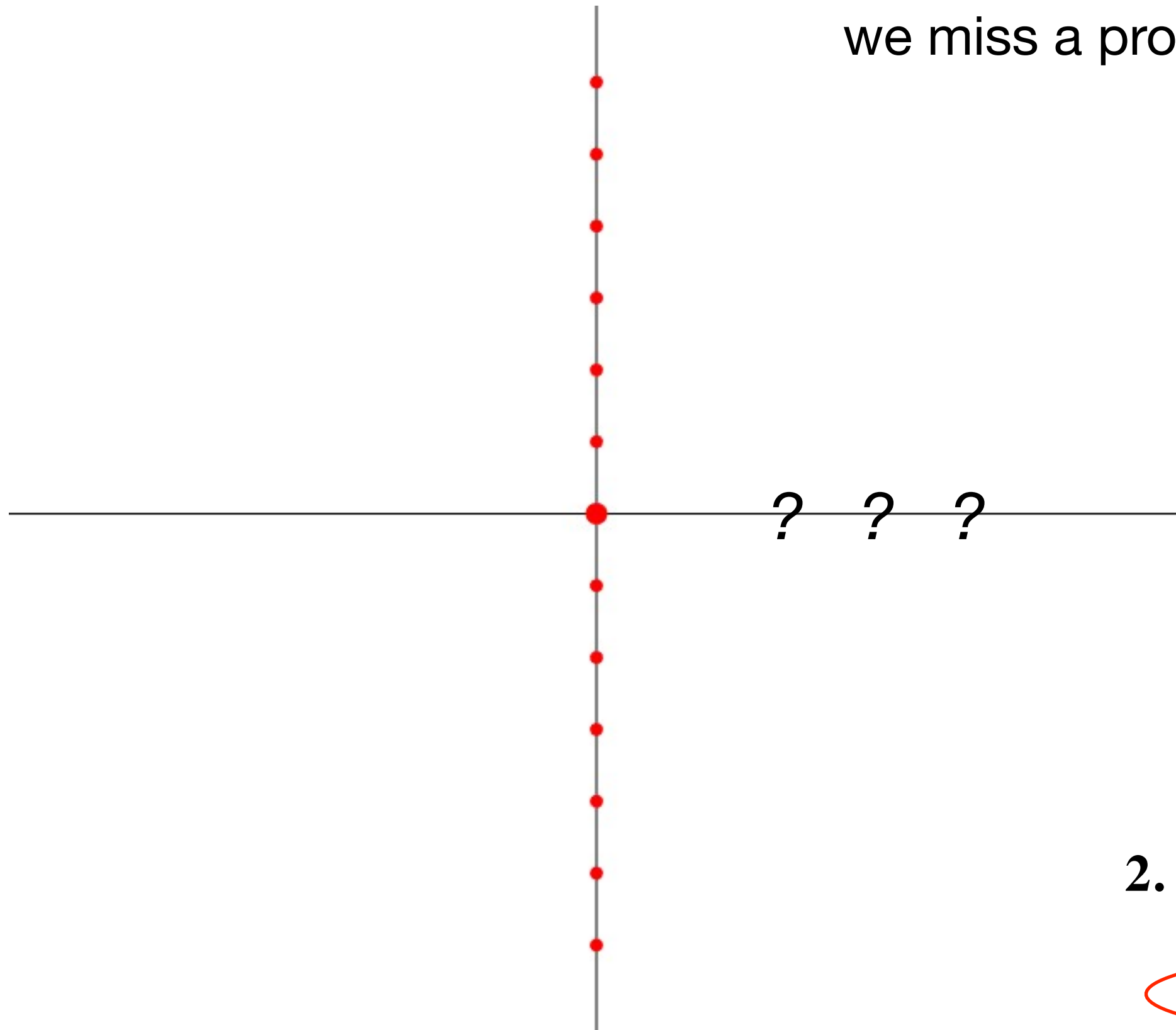


Complex chemical potential plane

SIGN PROBLEM for finite density Lattice QCD:

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Simulating QCD at finite density

Philippe de Forcrand*

Institute for Theoretical Physics, ETH Zürich, CH-8093 Zürich, Switzerland

and

CERN, Physics Department, TH Unit, CH-1211 Geneva 23, Switzerland

E-mail: forcrand@phys.ethz.ch

2. Sign problem

The sign problem is a necessary evil, unavoidable as soon as one integrates out the fermion fields and expresses the partition function in terms of the gauge fields. Analytic integration over each fermion species gives a factor $\det(\not{D} + m + \mu \gamma_0)$, where \not{D} is the massless Dirac operator and the last term appears when the chemical potential μ is non-zero. Now, \not{D} satisfies γ_5 -hermiticity: $\gamma_5 \not{D} \gamma_5 = \not{D}^\dagger$, so that

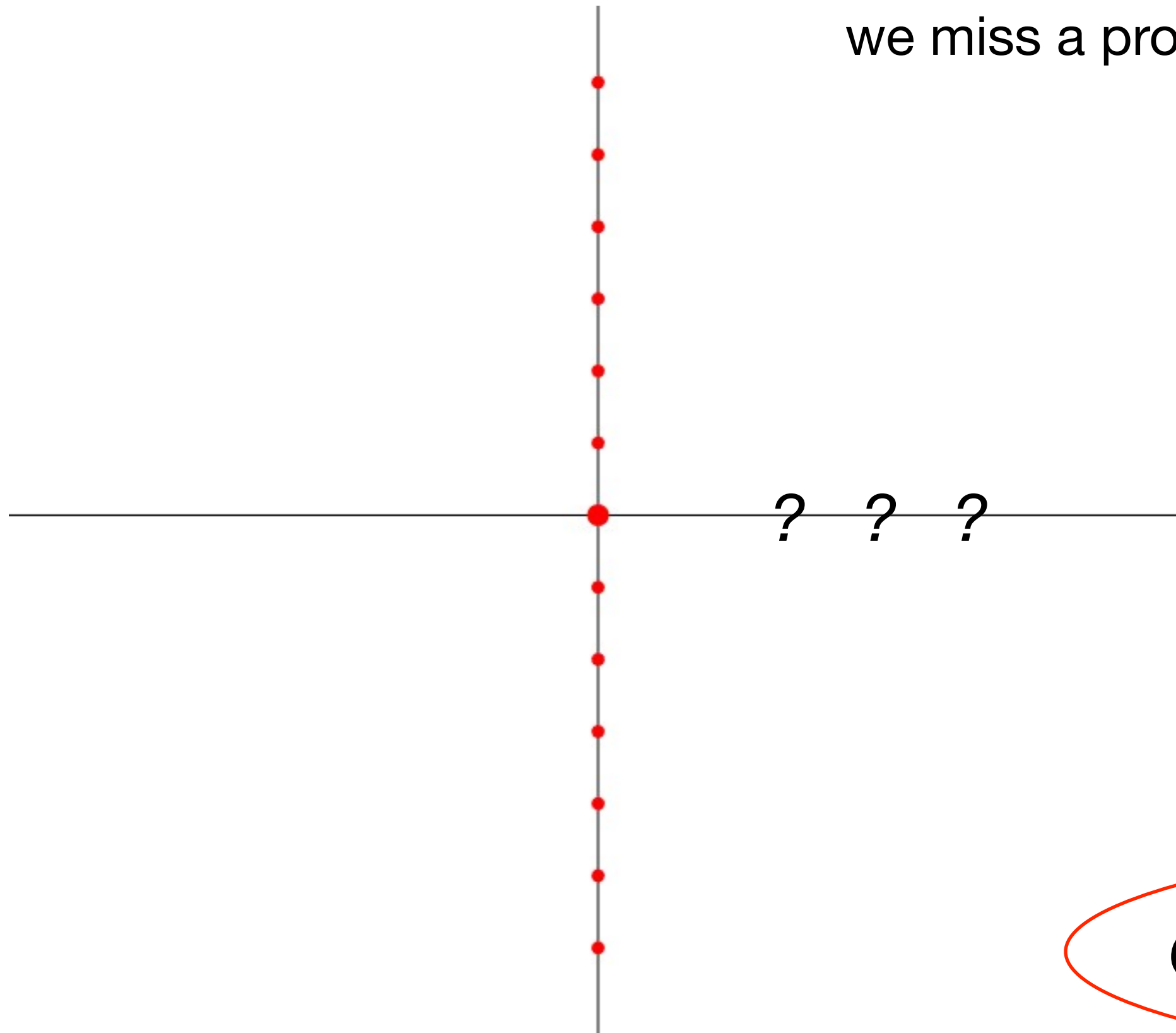
$$\gamma_5 (\not{D} + m + \mu \gamma_0) \gamma_5 = \not{D}^\dagger + m - \mu \gamma_0 = (\not{D} + m - \mu^* \gamma_0)^\dagger \quad (2.1)$$

Taking the determinant on both sides gives $\det(\not{D} + m + \mu \gamma_0) = \det^*(\not{D} + m - \mu^* \gamma_0)$, which constrains the determinant to be real only if μ is zero or pure imaginary.

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$$\gamma_5(\not{D} + m + \mu\gamma_0)\gamma_5 = \not{D}^\dagger + m - \mu\gamma_0 = (\not{D} + m - \mu^*\gamma_0)^\dagger$$

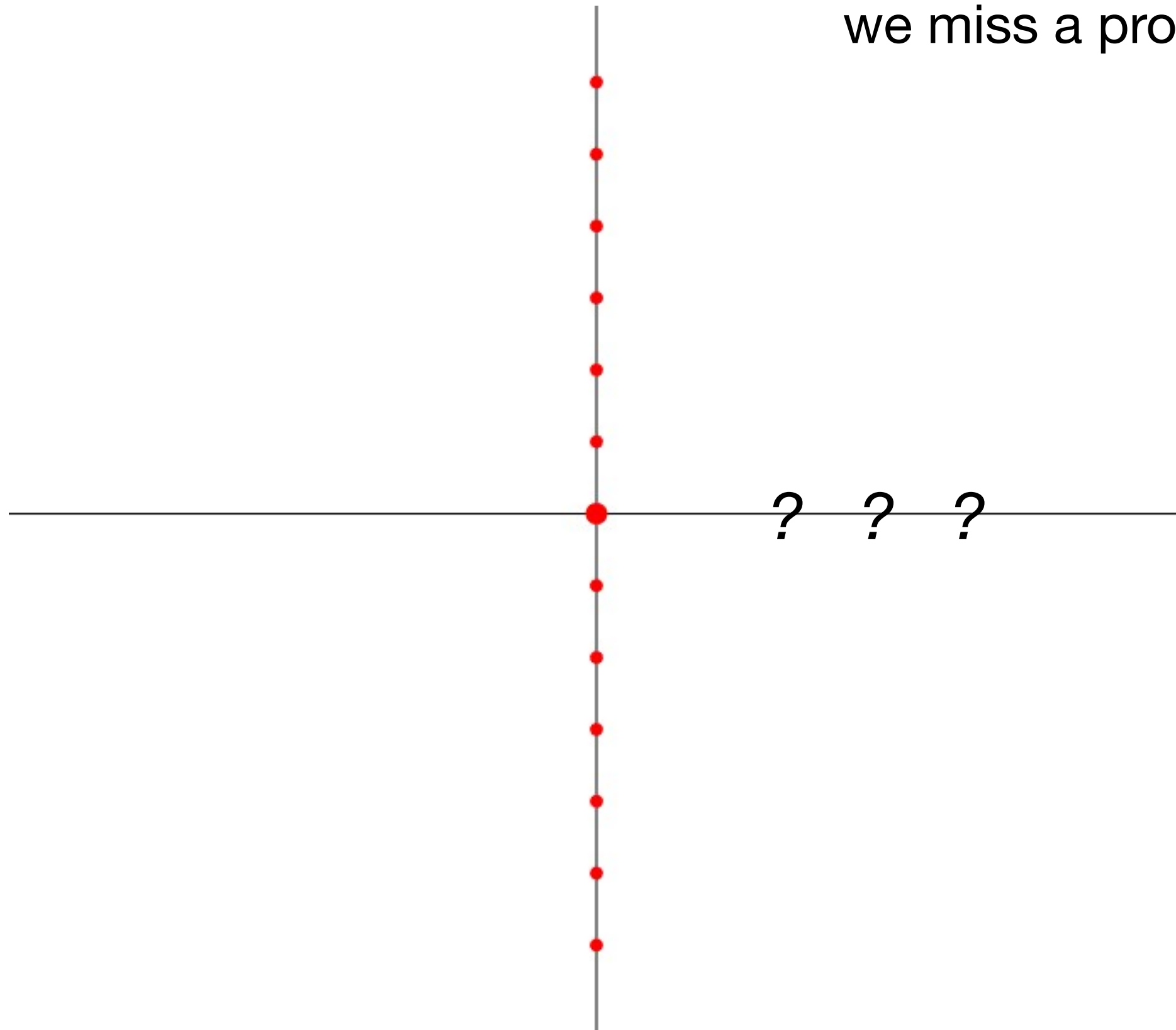
$$\det(\not{D} + m + \mu\gamma_0) = \det^*(\not{D} + m - \mu^*\gamma_0)$$

Fermionic determinant is **REAL** only for **ZERO** or **IMAGINARY** values of the chemical potential!

SIGN PROBLEM for **finite density Lattice QCD**:

we miss a properly defined (positive) measure in the path integral! ... no MC simulation

(... but everything is fine on the imaginary axis)



Mainly two working solutions:

- Compute **Taylor expansions** at $\mu_B = 0$
- Compute on the **imaginary axis** $\mu_B = i\mu_I$

The two solutions are obviously related ... and both imply (strictly speaking) an **ANALYTIC CONTINUATION**

There are tensions in between different results for Taylor coefficients in the literature...

Agenda

- An invitation (1. **sign problem...**)
- An invitation (2. analytic continuation from **multi-point Padé**)
- The sign problem as an **inverse problem ...**
- ... and something more on **other inverse problems ...**
- WHAT WE CAN DO WITH ALL THIS ...

Until sometime ago, if you had asked me about analytic continuation, I would have told you a few words on multi-point PADÈ

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Suppose you know the **values** of a **function** (and of its derivatives) at a number of points

$$\dots, f(z_k), f'(z_k), \dots, f^{(s-1)}(z_k), \dots \quad k = 1 \dots N$$

If you want to **approximate the function with a rational function**

$$R_n^m(z) = \frac{P_m(z)}{\tilde{Q}_n(z)} = \frac{P_m(z)}{1 + Q_n(z)} = \frac{\sum_{i=0}^m a_i z^i}{1 + \sum_{j=1}^n b_j z^j}$$

the obvious requirement is that

$$R_n^{m(j)}(z_k) = f^{(j)}(z_k) \quad k = 1 \dots N, \quad j = 0 \dots s - 1$$

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the obvious requirement is that

$$R_n^m{}^{(j)}(z_k) = f^{(j)}(z_k) \quad k = 1 \dots N, \quad j = 0 \dots s - 1$$

This is the starting point for a **multi-point Padè approximation**: solve the linear system

$$\begin{aligned} & \dots \\ & P_m(z_k) - f(z_k)Q_n(z_k) = f(z_k) \\ & P'_m(z_k) - f'(z_k)Q_n(z_k) - f(z_k)Q'_n(z_k) = f'(z_k) \\ & \dots \end{aligned}$$

from which we want to get the unknown

$$\{a_i \mid i = 0 \dots m\} \quad \{b_j \mid j = 1 \dots n\} \quad n + m + 1 = N s$$

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Any useful ...?

Yes! LATTICE QCD at **IMAGINARY** values of the **baryonic chemical potential**

PHYSICAL REVIEW D **105**, 034513 (2022)

Contribution to understanding the phase structure of strong interaction matter: Lee-Yang edge singularities from lattice QCD

P. Dimopoulos¹, L. Dini,² F. Di Renzo¹, J. Goswami², G. Nicotra², C. Schmidt²,
S. Singh^{1,*}, K. Zambello¹ and F. Ziesché²

... where we computed and “multi-point Padè approximated”

$$\begin{aligned} \chi_n^B(T, V, \mu_B) &= \left(\frac{\partial}{\partial \hat{\mu}_B} \right)^n \frac{\ln Z(T, V, \mu_l, \mu_s)}{VT^3} \\ &= \left(\frac{1}{3} \frac{\partial}{\partial \hat{\mu}_l} + \frac{1}{3} \frac{\partial}{\partial \hat{\mu}_s} \right)^n \frac{\ln Z(T, V, \mu_l, \mu_s)}{VT^3} \end{aligned}$$

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Any useful ...?

Yes! LATTICE QCD at **IMAGINARY** values of the **baryonic chemical potential**

... a natural analytic continuation to real chemical potential!

... and not only that: **singularities!**

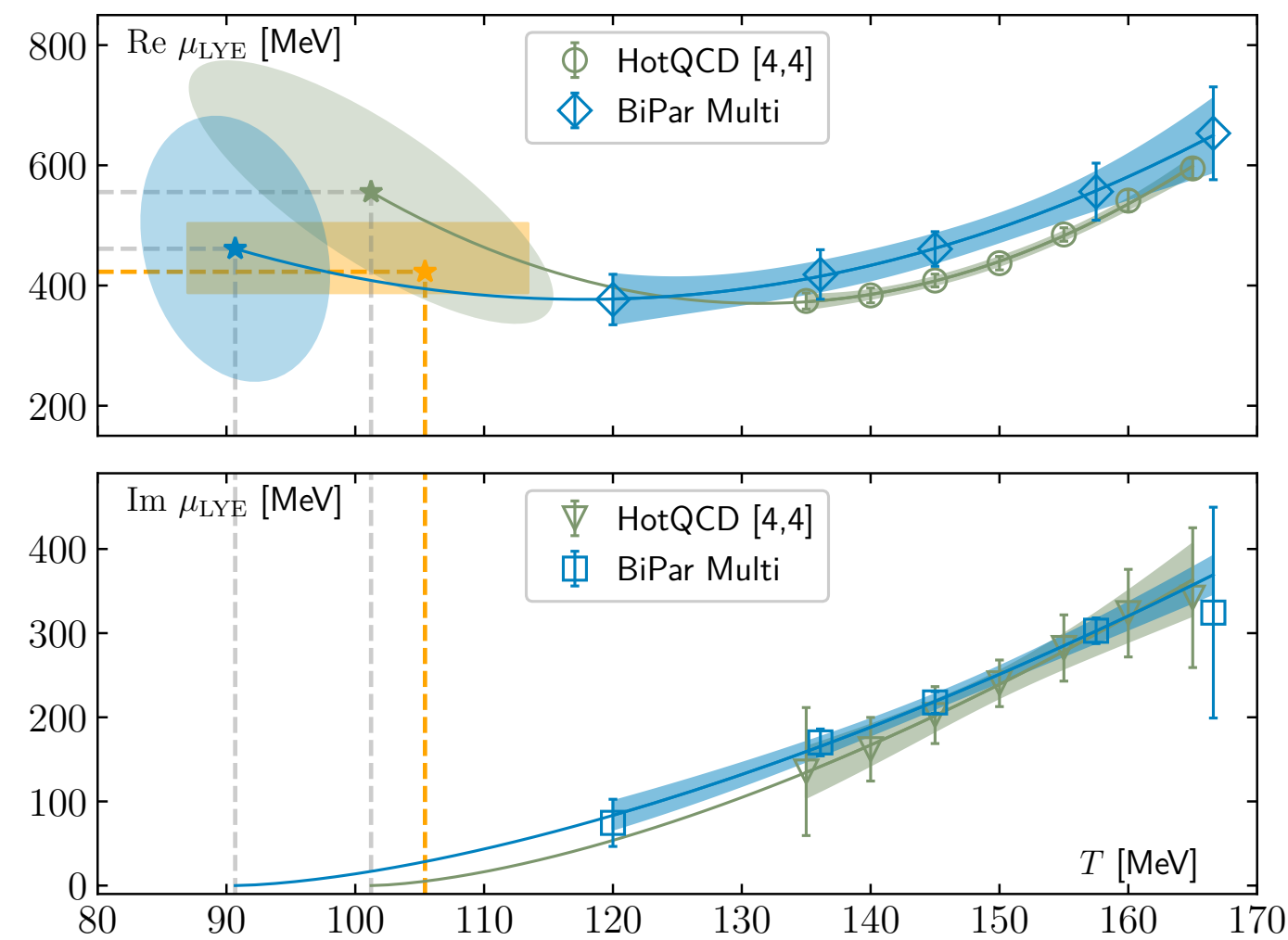
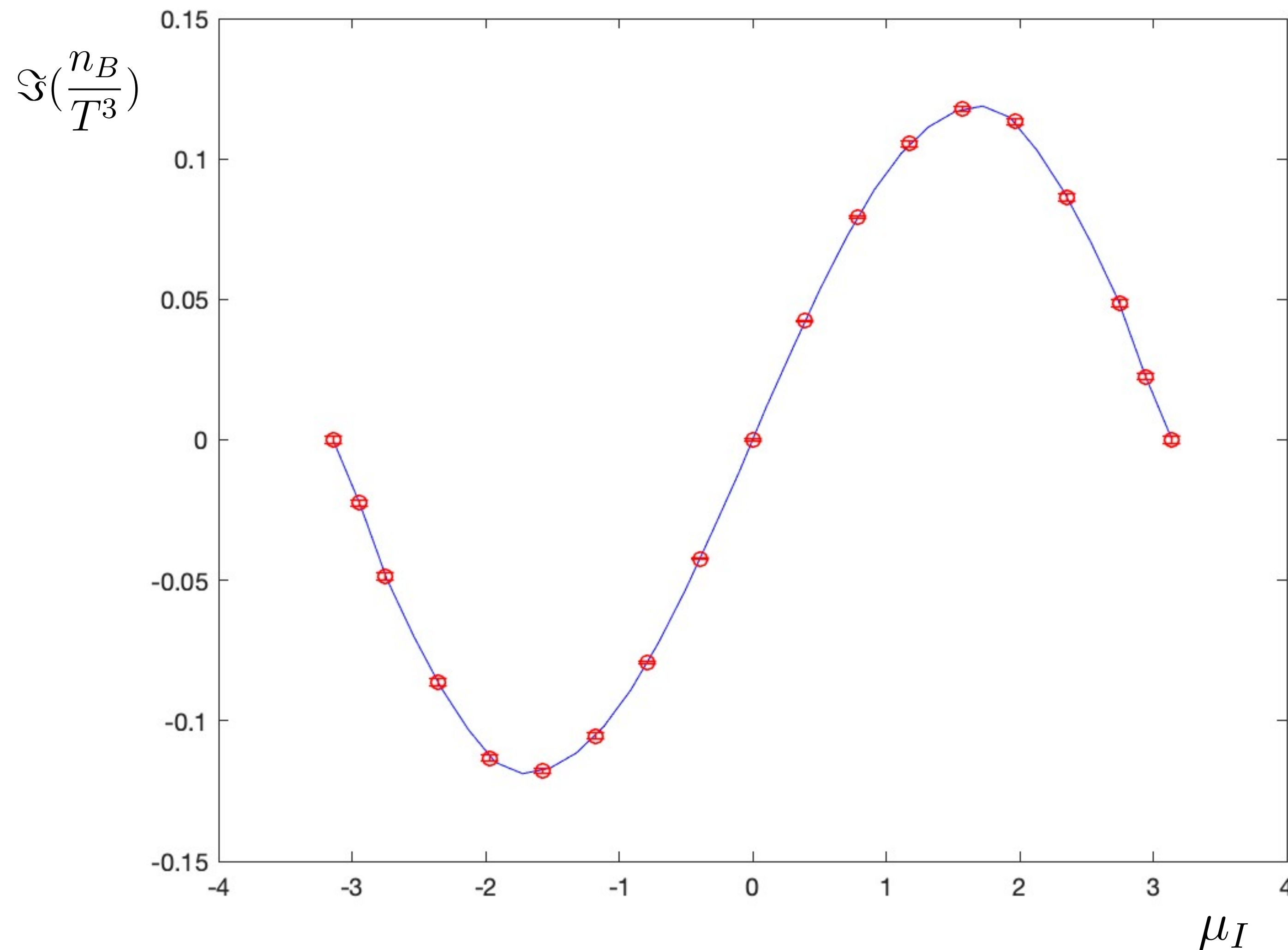


FIG. 4. Scaling fits for the LYE singularities related to the CEP. Green data come from a [4,4] Padé from Ref. [7]. Blue data come from the multi-point Padé. *Top*: Scaling of the real part. *Bottom*: Scaling of the imaginary part. The ellipses shown in the top panel represent the 68% confidence region deduced from the covariance matrix of the fit. The orange box indicates the AIC weighted estimate (6).

In the following we will play with a few **Bielefeld Parma Collaboration data**
2+1 HISQ at physical quarks mass, at fixed cutoff ($N_\tau = 6$)

...so, let's look at analytic continuation of our PADÉ approximant

$T = 157.5 (\sim 155) \text{ MeV}$



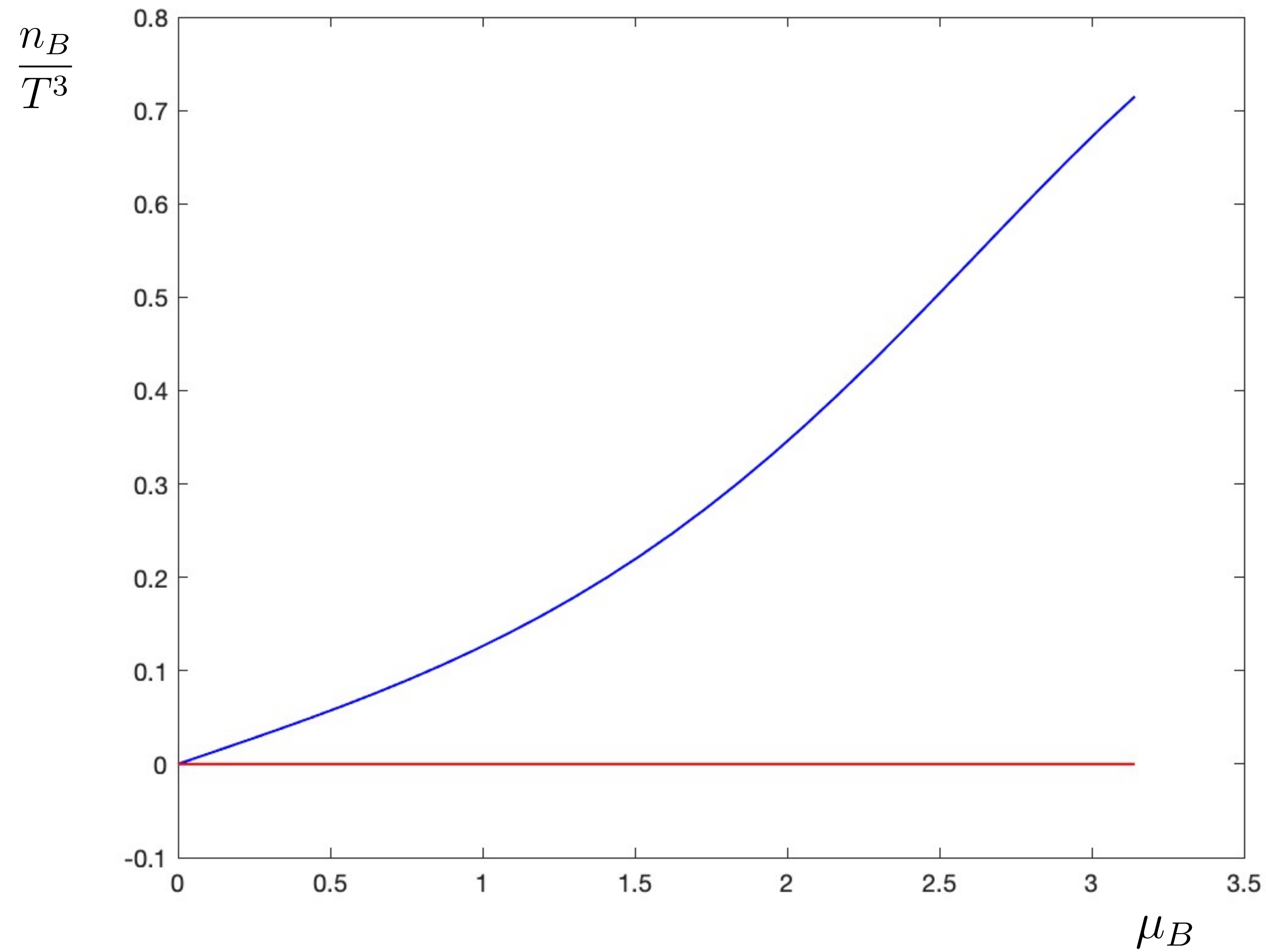
... which is pretty simple (we will be concerned with the **number density**):

you take your rational function, which describes very well data at **IMAGINARY VALUES** of μ_B

CAVEAT: errors on data points are there ... no error shown on the interpolating function (*as for now ...*)

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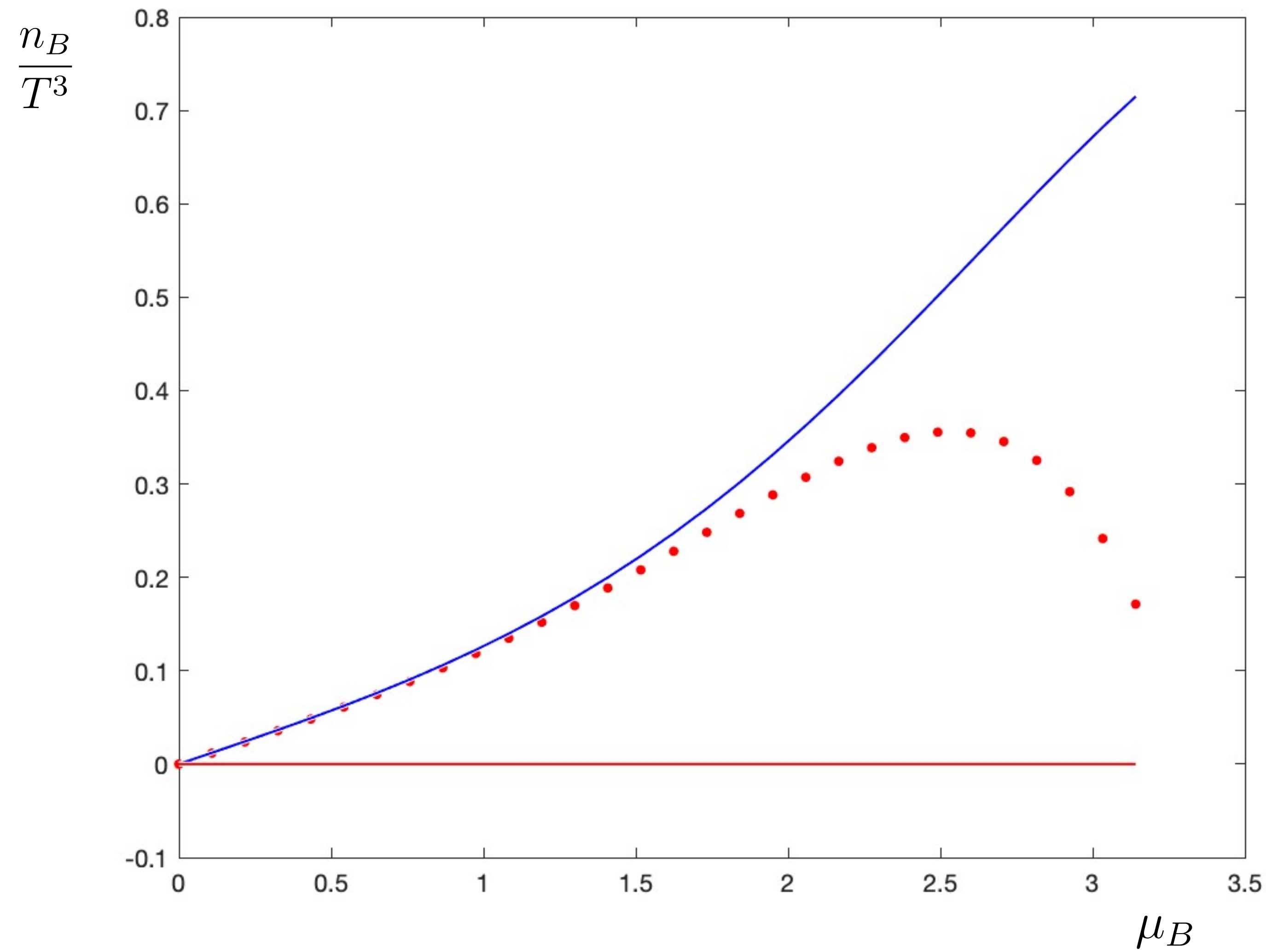
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... and you simply compute it for **REAL VALUES** of μ_B

CAVEAT: no error shown as for now ... here we are concerned with *trends*... **FIXED CUTOFF!**

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you take your rational function, which describes very well data at **IMAGINARY VALUES** of μ_B

... and you simply compute it for **REAL VALUES** of μ_B

You can compare the result with HotQCD results

PHYSICAL REVIEW D **105**, 074511 (2022)

Taylor expansions and Padé approximants for cumulants of conserved charge fluctuations at nonvanishing chemical potentials

D. Bollweg¹, J. Goswami², O. Kaczmarek², F. Karsch², Swagato Mukherjee³, P. Petreczky³, C. Schmidt² and P. Scior³

(HotQCD Collaboration)

CAVEAT: no error shown as for now ... here we are concerned with *trends*... **FIXED CUTOFF!**

Finite density QCD as an **inverse problem**

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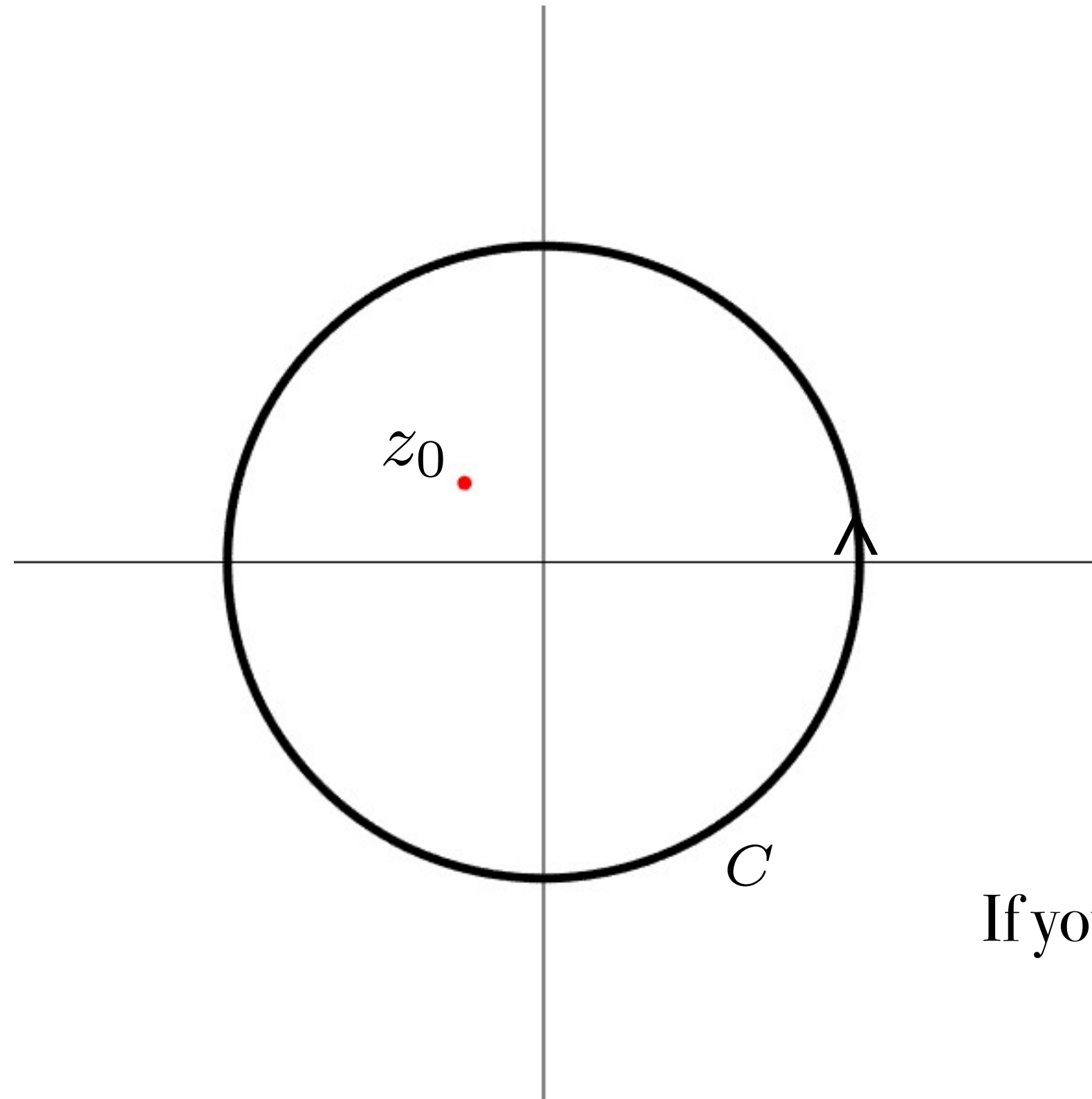
... *aka* How to trade a difficult problem for another (even more?) difficult one ...

What does **ANALYTICITY** mean? ... (analytic functions aka **holomorphic**...)

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One simple way of thinking of it is that *you can perfectly know such functions from an apparently limited amount of information.*

What does **ANALYTICITY** mean? ... (analytic functions aka **olomorphic**...)

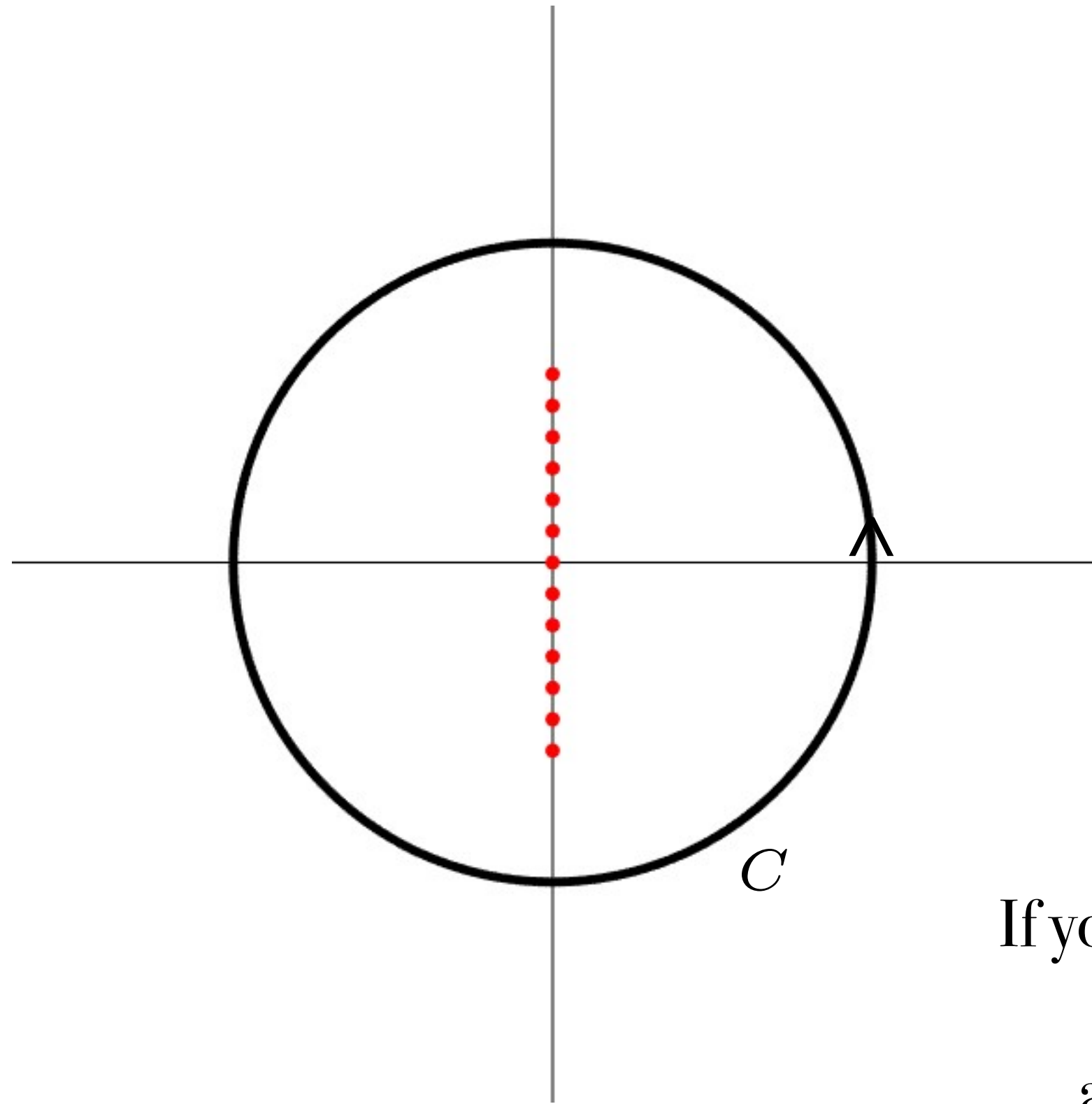


CAUCHY FORMULA

$$f(z_0) = \frac{1}{2\pi i} \oint_C \frac{f(z)}{z - z_0} dz$$

If you know the function on the contour, you can compute it at any point inside... sounds good!

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CAUCHY FORMULA

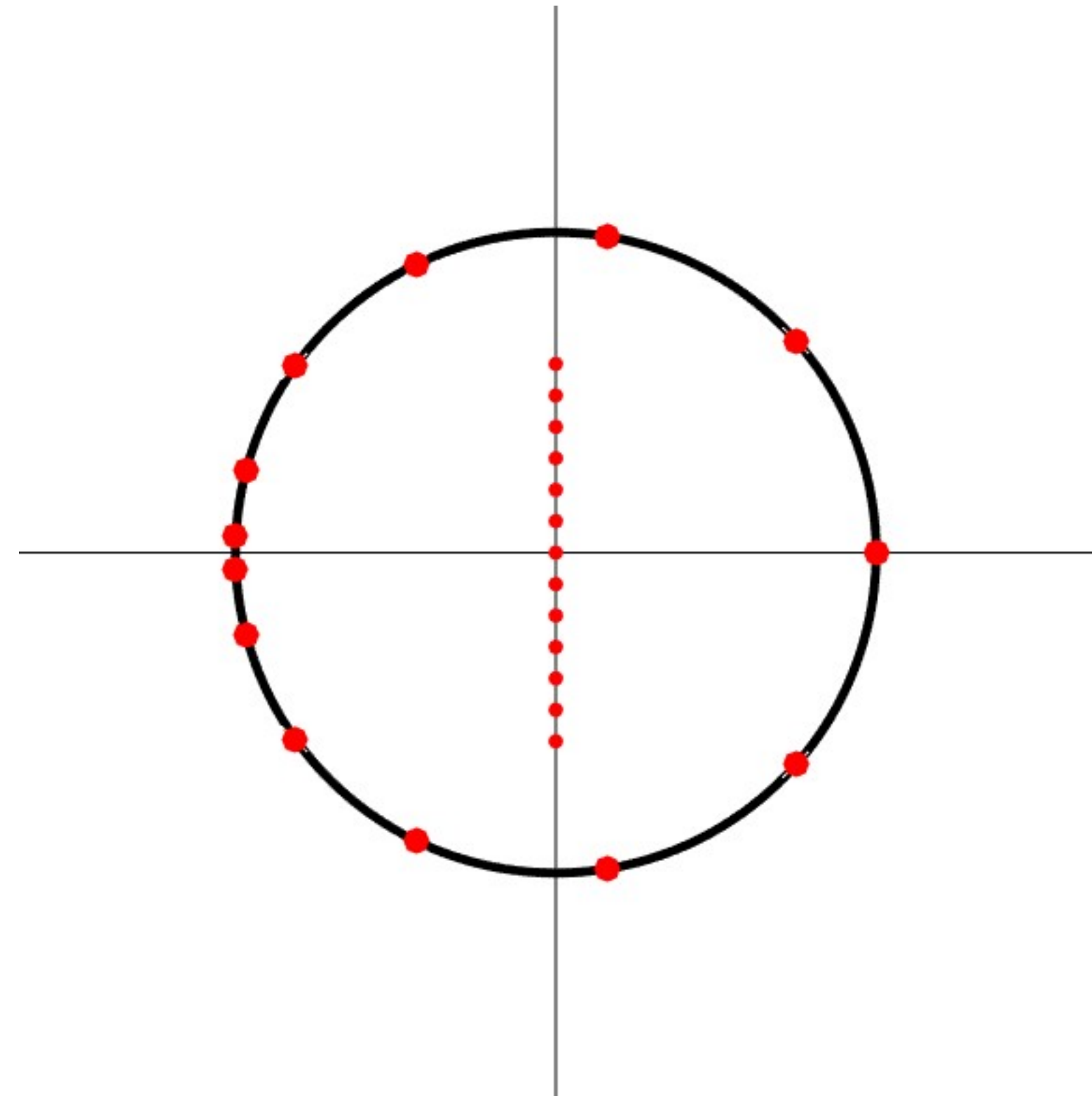
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If you know the function on the contour, you can compute it at any point inside... sounds good!

... at any point, including the (only) ones we can compute (on the imaginary axis) in our case...

With your favourite **QUADRATURE** method ... you can go **numeric**!

De facto, you would like to think of **Legendre quadrature**



$$f(z_0) = \frac{1}{2\pi} \int_0^{2\pi} \frac{f(R e^{i\theta}) R e^{i\theta}}{R e^{i\theta} - z_0} d\theta \simeq \frac{1}{2\pi} \sum_{k=1}^n w_k \frac{f(R e^{i\theta_k}) R e^{i\theta_k}}{R e^{i\theta_k} - z_0}$$

$$y_i = \frac{1}{2\pi} \sum_{k=1}^n w_k \frac{R e^{i\theta_k}}{R e^{i\theta_k} - z_i} \hat{f}_k, \quad i = 1, 2, \dots, n$$

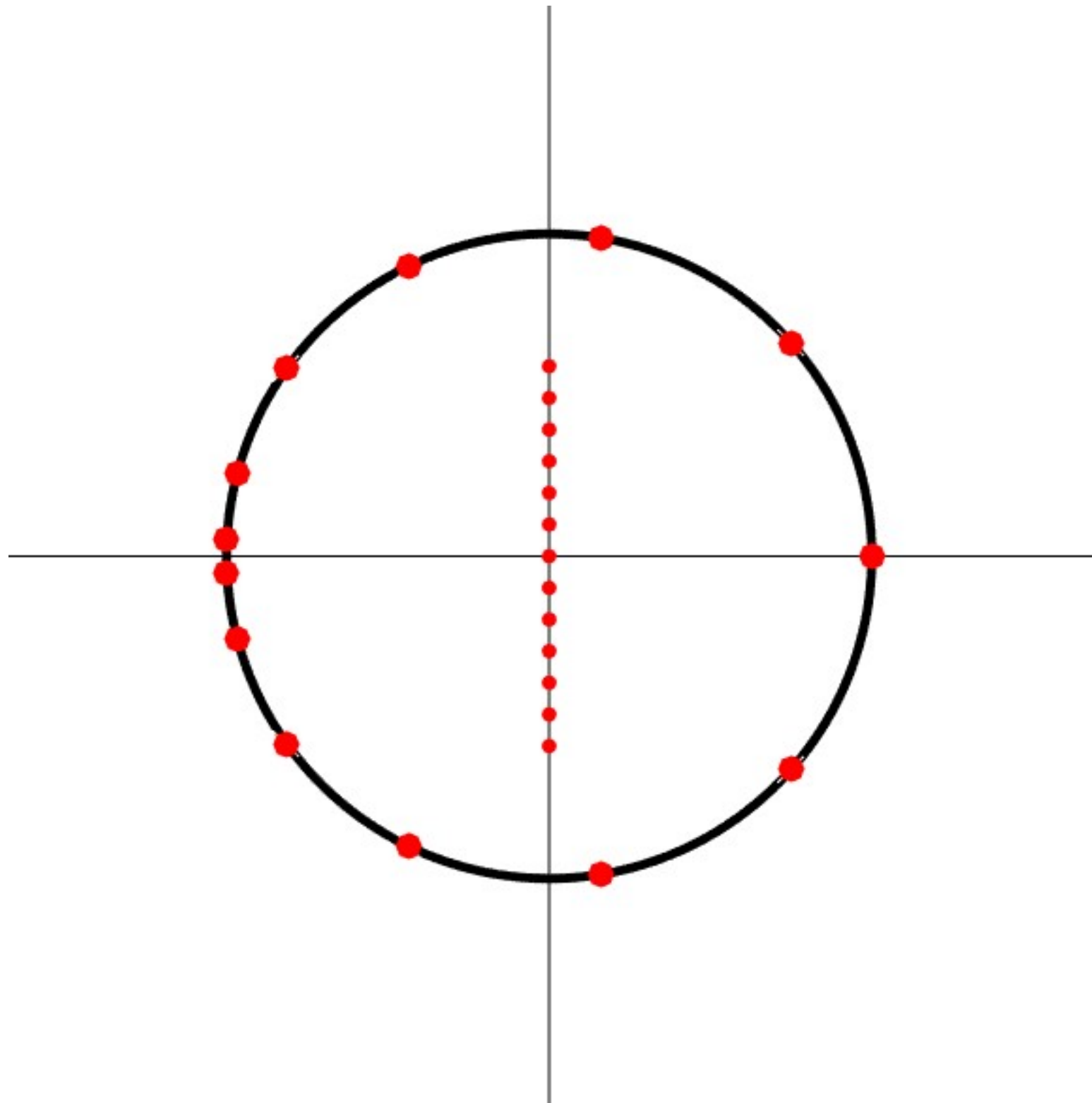
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... and then you are ready for your (BRAVE) **INVERSE PROBLEM!**



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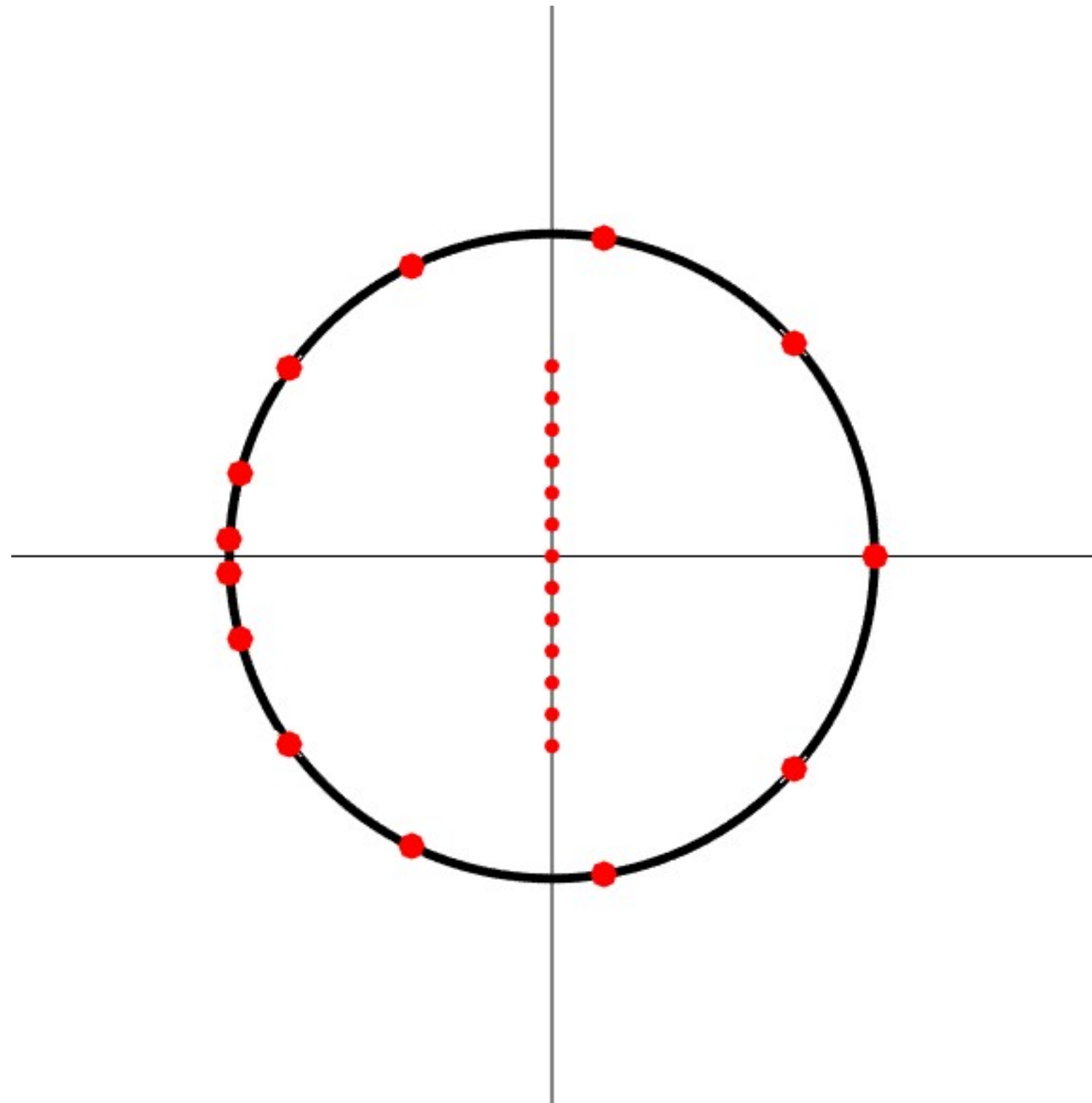
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... and then you are ready for your (BRAVE) **INVERSE PROBLEM**!

$$A_{ik} = \frac{1}{2\pi} w_k \frac{R e^{i\theta_k}}{R e^{i\theta_k} - z_i}$$

$$A \mathbf{x} = \mathbf{b}$$

SOLVE for the \hat{f}_k !



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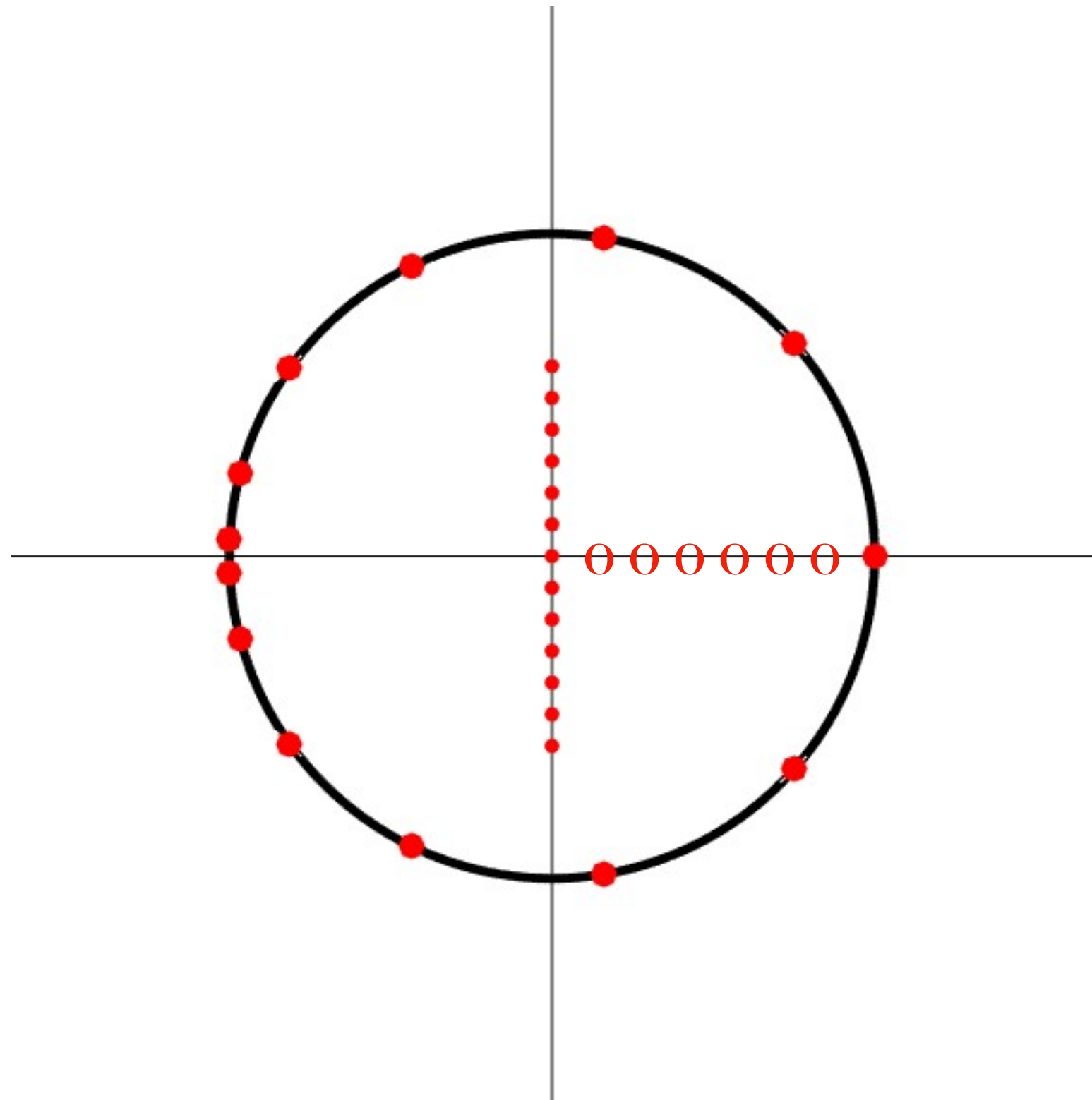
... and when you have solved your (BRAVE) **INVERSE PROBLEM(!)** ...

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$$A \mathbf{x} = \mathbf{b}$$

SOLVE for the \hat{f}_k !

Once you have the values on the contour, **you can compute on the REAL AXIS!** (o o o o o o)



With your favourite **QUADRATURE** method ... you can go **numeric**!

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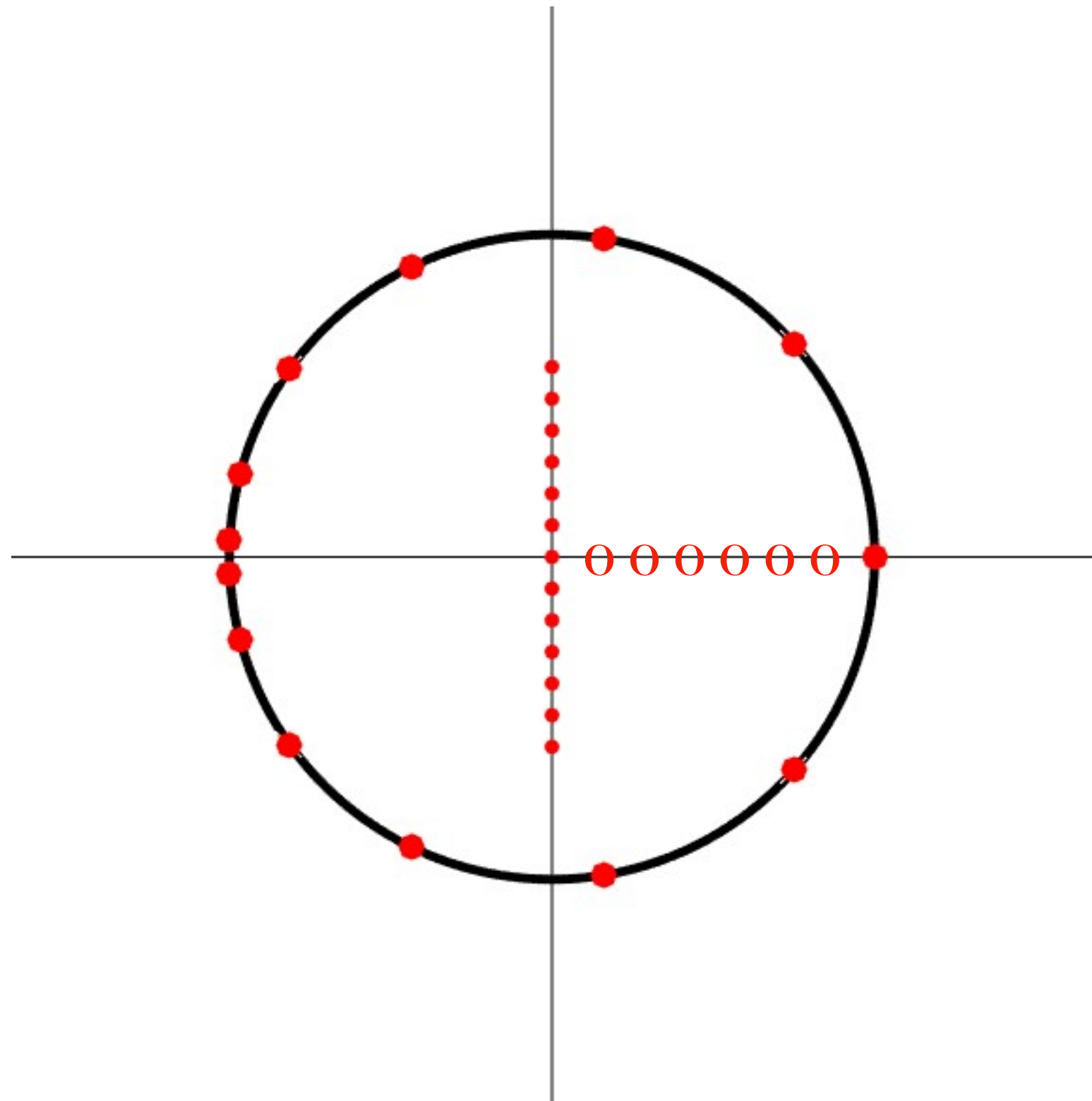
... a BRAVE **INVERSE PROBLEM**!...

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$$A \mathbf{x} = \mathbf{b}$$

SOLVE for the \hat{f}_k !

A very **SIMPLE** idea! ... a **NAIVE** one ... **INVERSE PROBLEM!!!**



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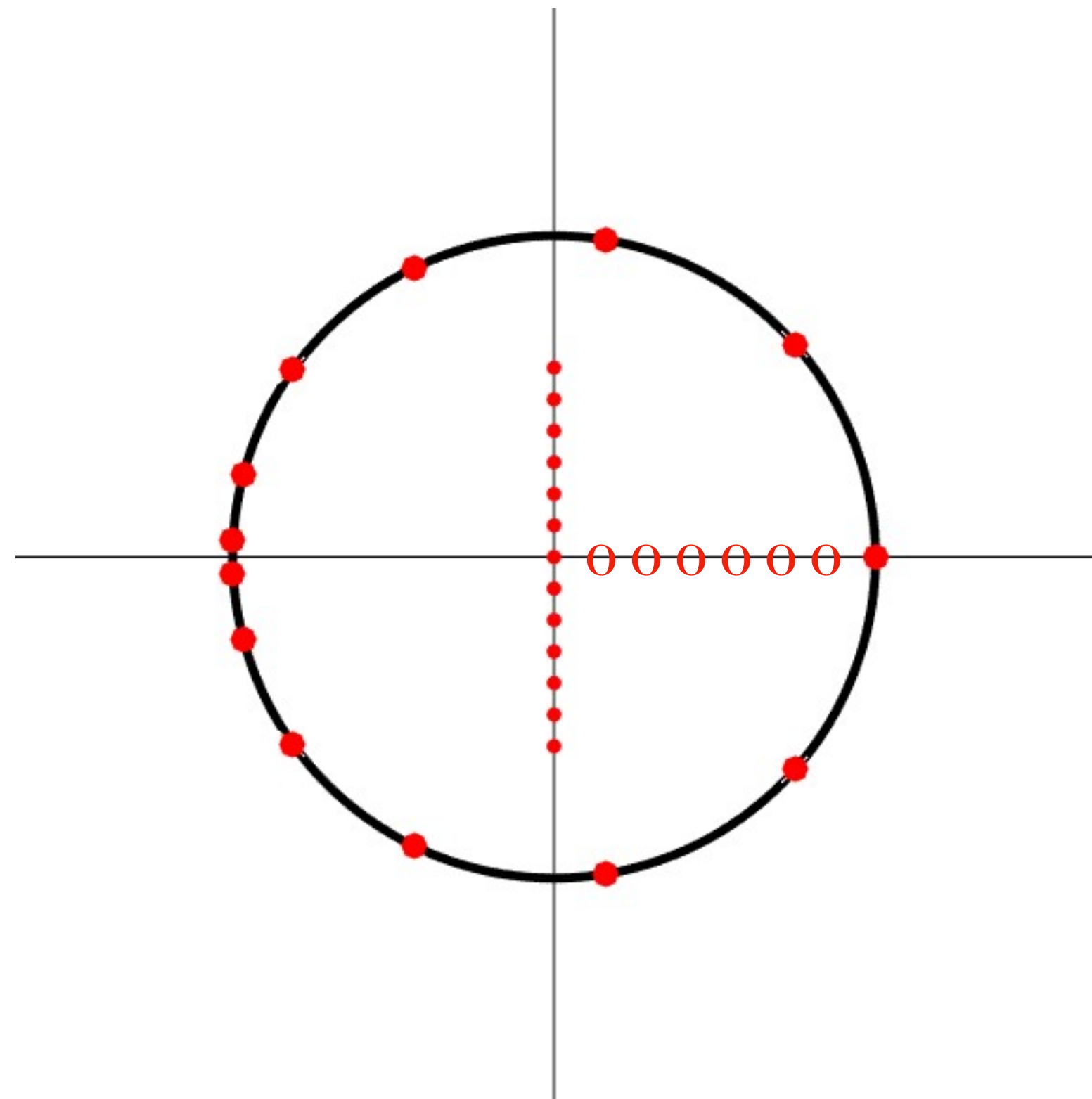
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SOLVE for the \hat{f}_k !



Can we hope it could work? ... are we afraid it **should not**?

... and it **could(?) / should(?) not** for a combination of

- (a) bad **condition number** of the linear system
- (b) the quadrature formula being **NOT exact**

Much care is needed ... and so we will perform some tests...

With your favourite **QUADRATURE** method ... you can go **numeric**!

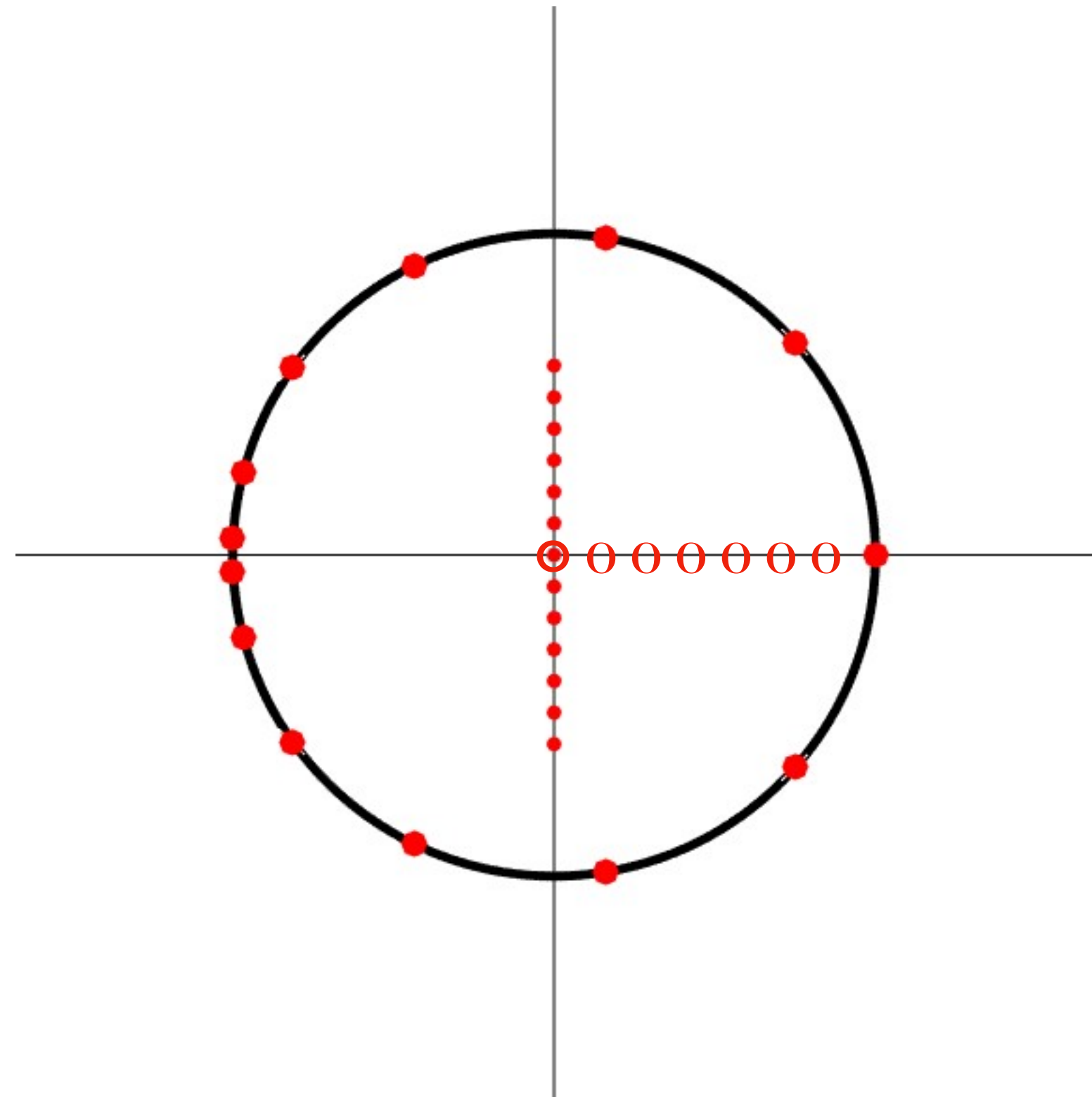
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SOLVE for the \hat{f}_k !



Not the end of the story ... Obviously, it is a good (CHEAP) idea to also remember **Cauchy formula for derivatives**

$$f^{(n)}(z_0) = \frac{n!}{2\pi i} \oint_C \frac{f(z)}{(z - z_0)^{n+1}} dz = \frac{n!}{2\pi i} \int_0^{2\pi} \frac{f(R \exp(i\theta)) R \exp(i\theta)}{(R \exp i\theta - z_0)^{n+1}} d\theta$$

LOOK! ... we tackled a funny inverse problem, a non-standard one ... other are very much investigated!

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SOMETHING ELSE you can do with the inverse problem machinery:

we can play the same game for inverse Laplace transform ...

You know this ...

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$$f(s) = \int_0^{\infty} e^{-ts} F(t) dt$$

You want this ...

We slightly rephrase the problem ...

$$f(s) = \int_0^{\infty} e^{-t} e^{-t(s-1)} F(t) dt$$

... and we can play the same game ...

$$f(s) = \int_0^{\infty} e^{-t} e^{-t(s-1)} F(t) dt \sim \sum_j w_j e^{-t_j(s-1)} F(t_j)$$

This time, Laguerre quadratures ...

... A brave attempt, but with a rational ...

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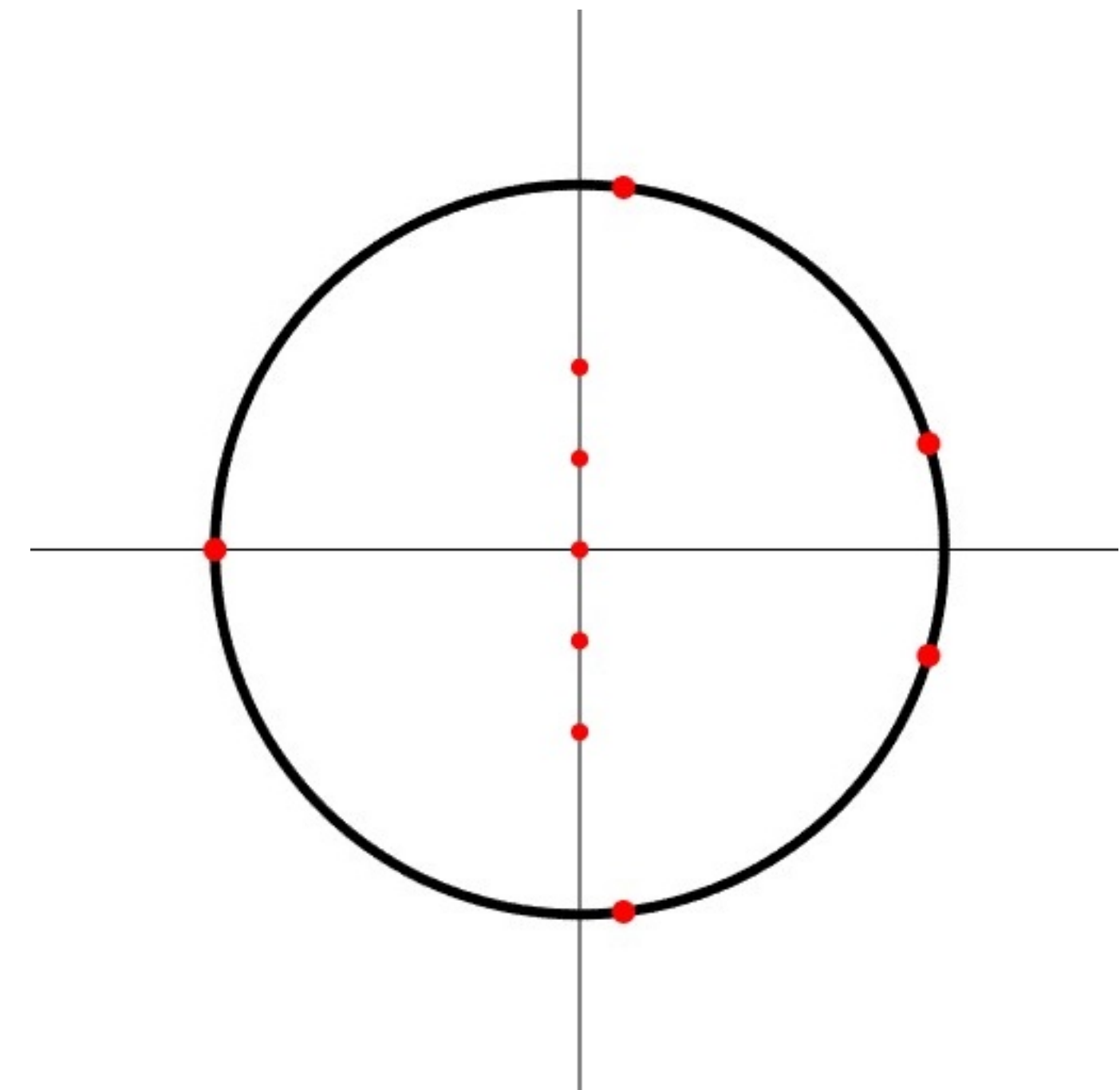
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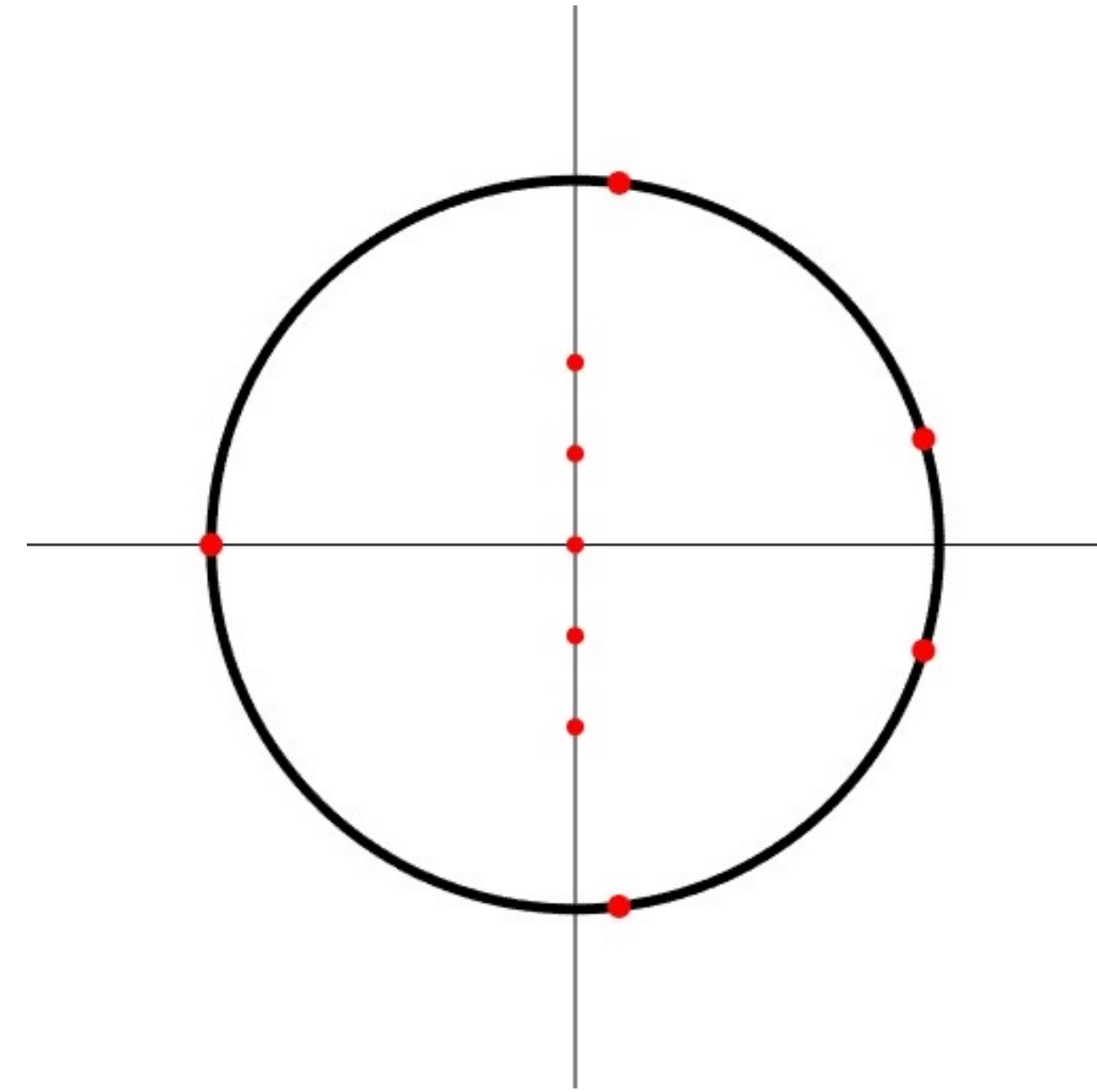
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Actually, both for (anti)Laplace and for the Cauchy formula,
we provide n values and try to get n as well ... e.g.



... A brave attempt, but with a rational ...

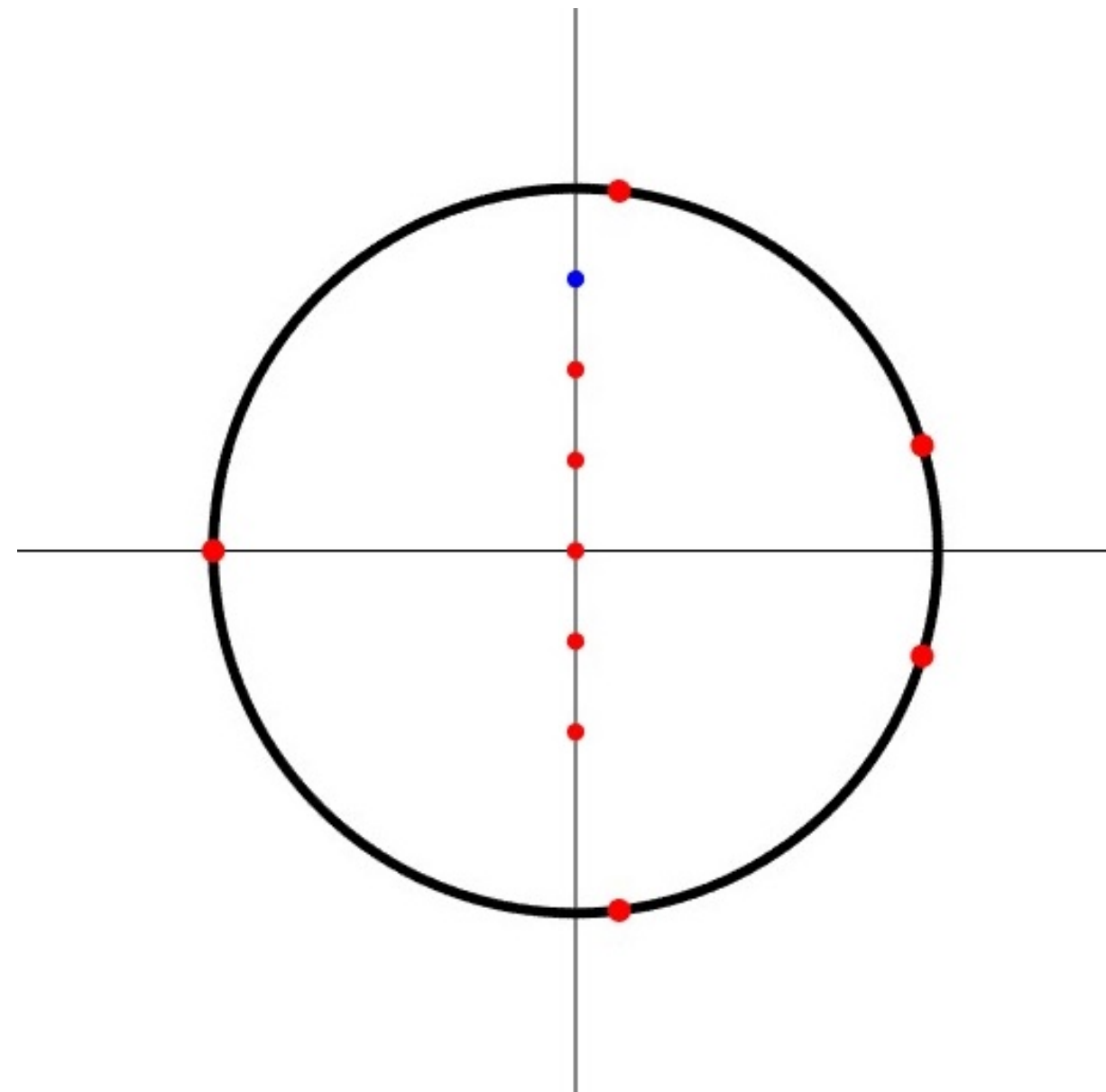
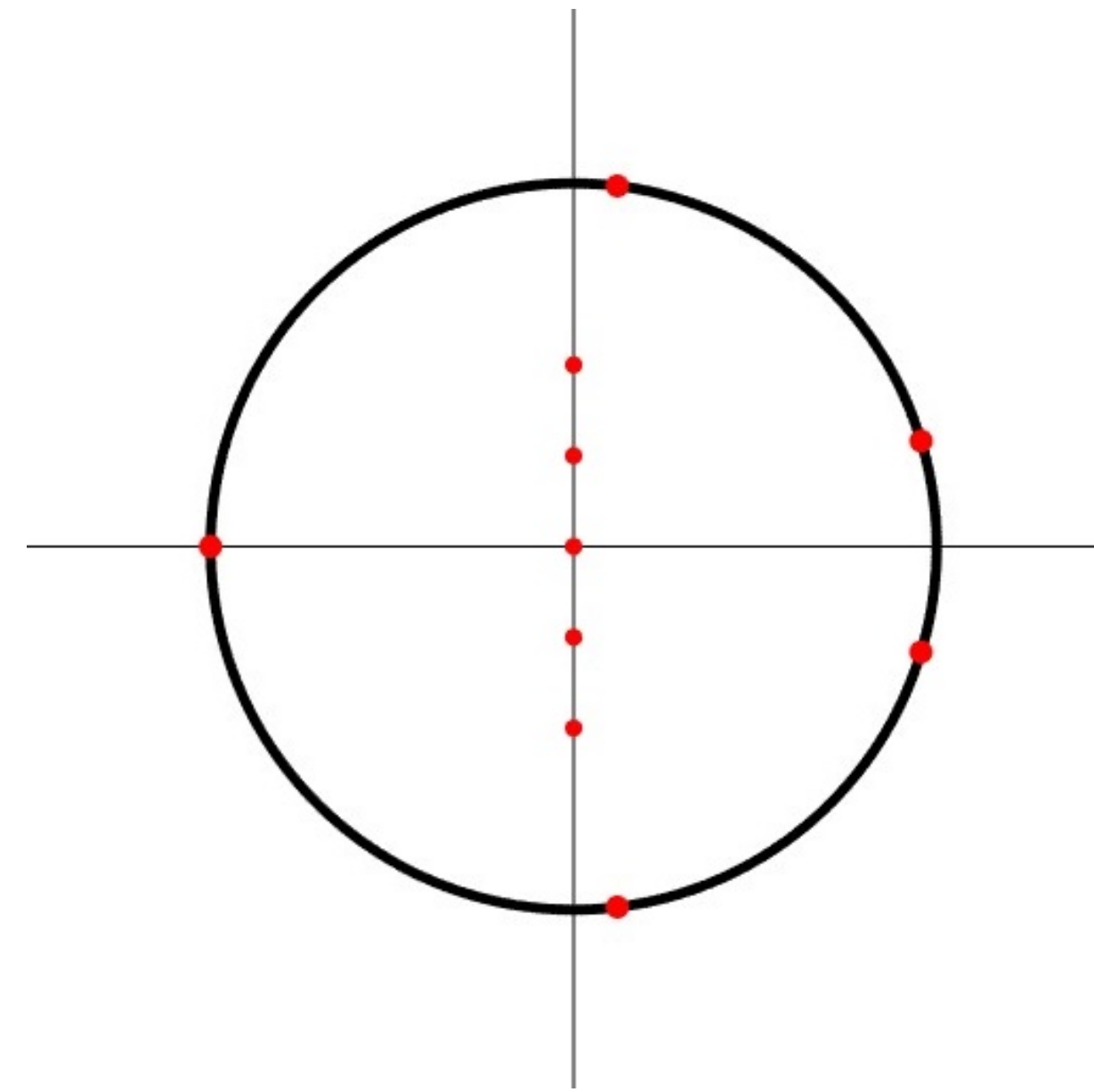
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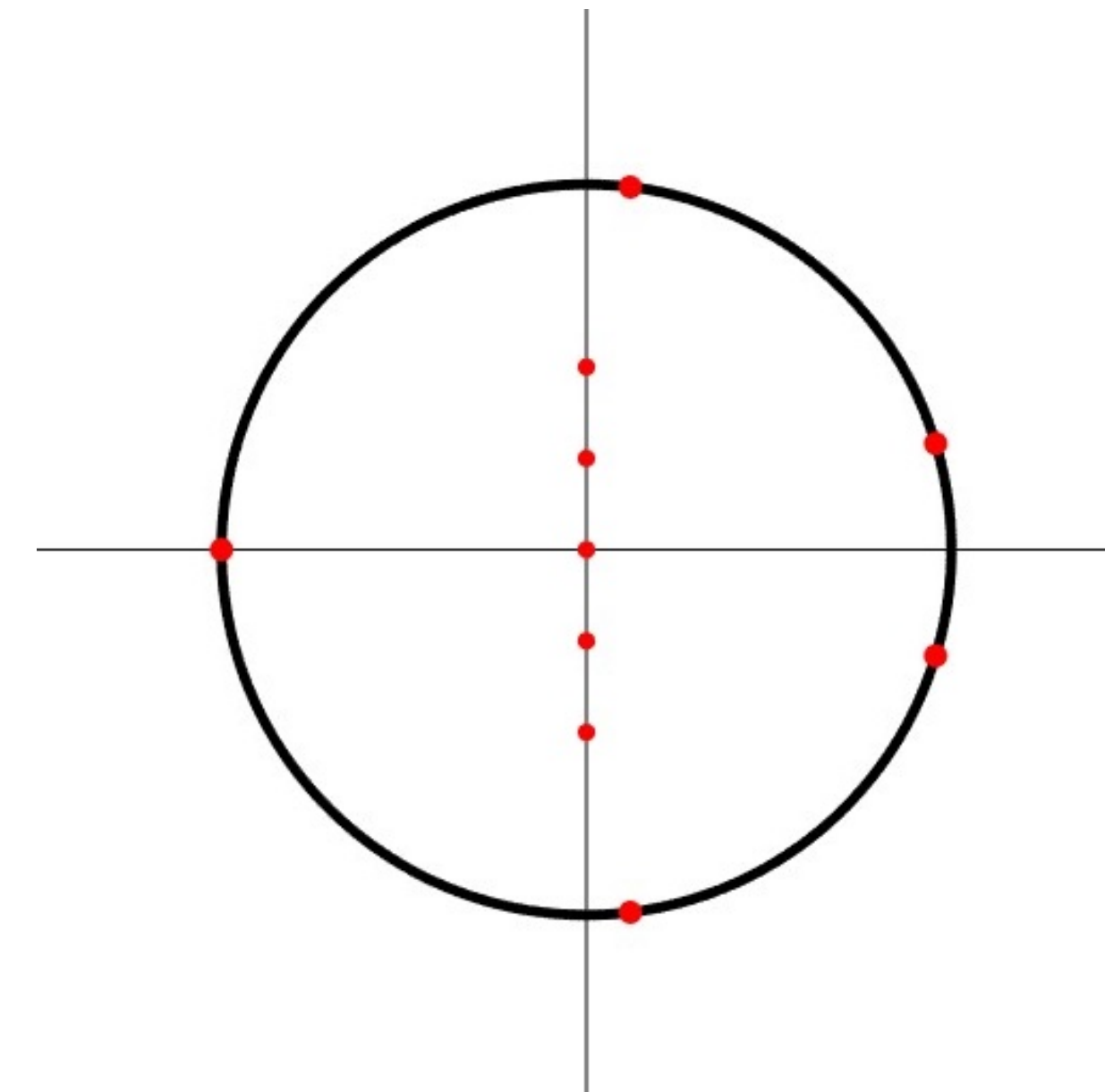
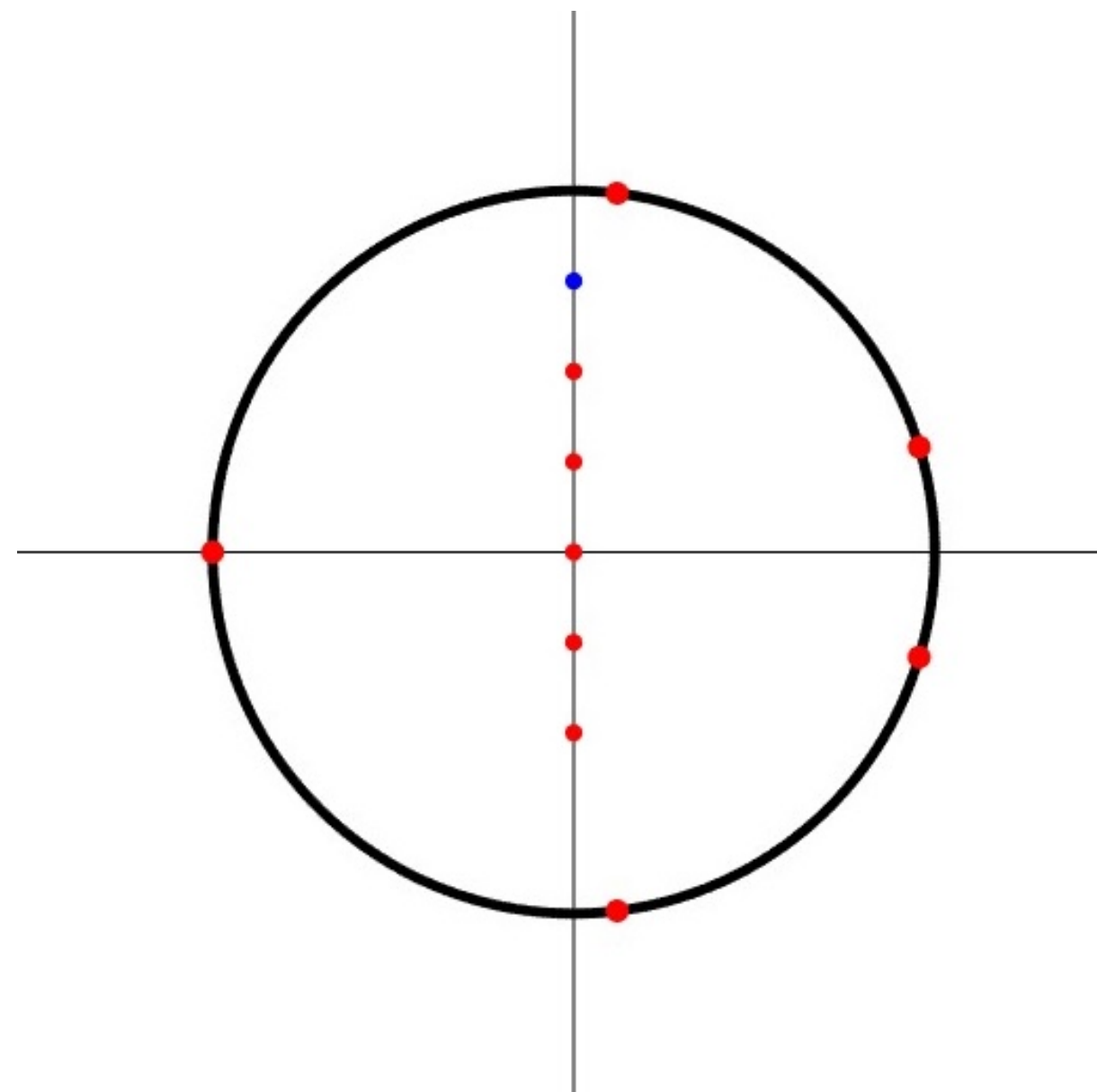
... but then ... simply add **one** (single) point ...



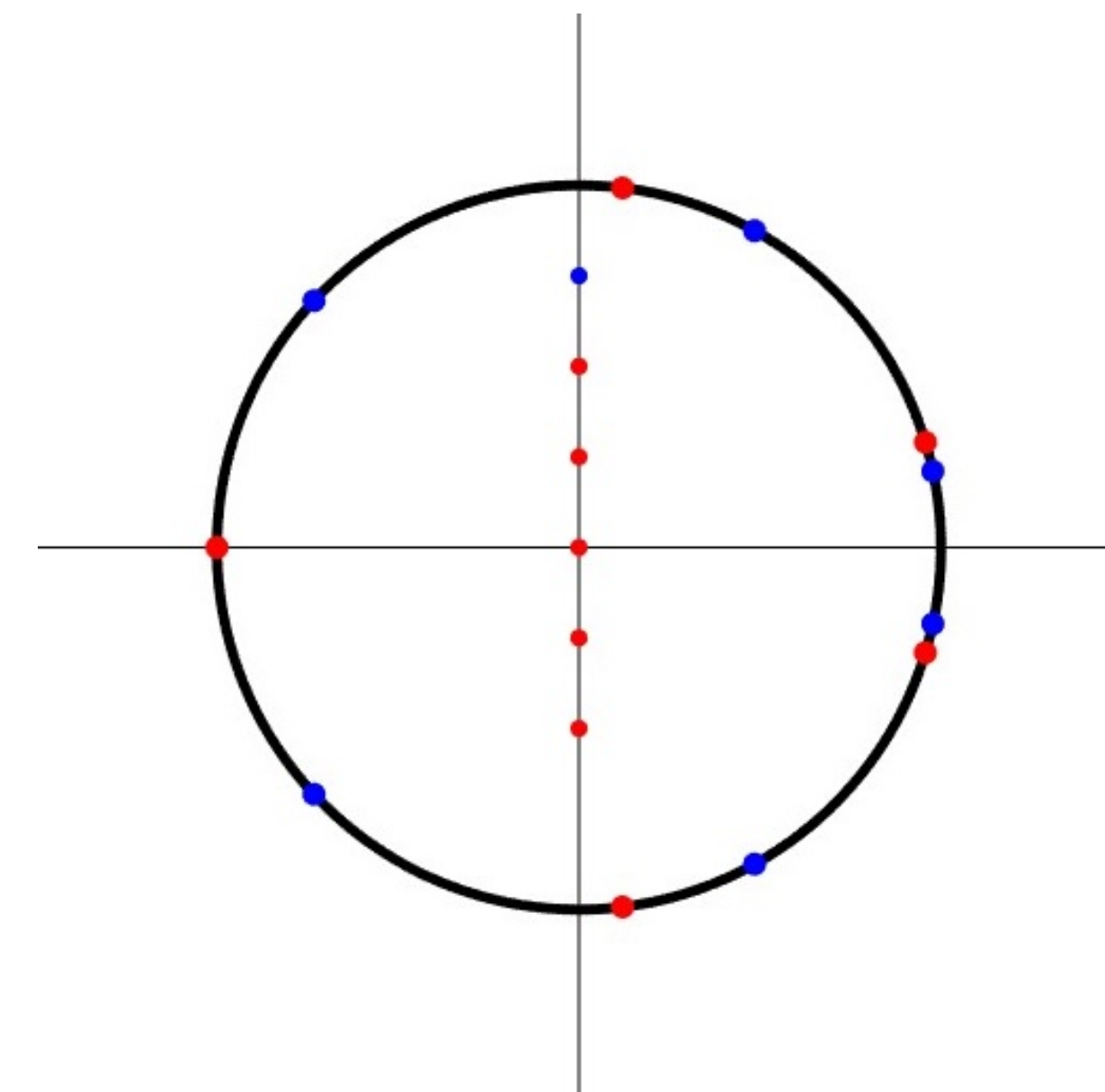
... A brave attempt, but with a rational ...

Actually, both for (anti)Laplace and for the Cauchy formula,
we provide n values and try to get n as well ... e.g.

... but then ... simply add **one** (single) point ...

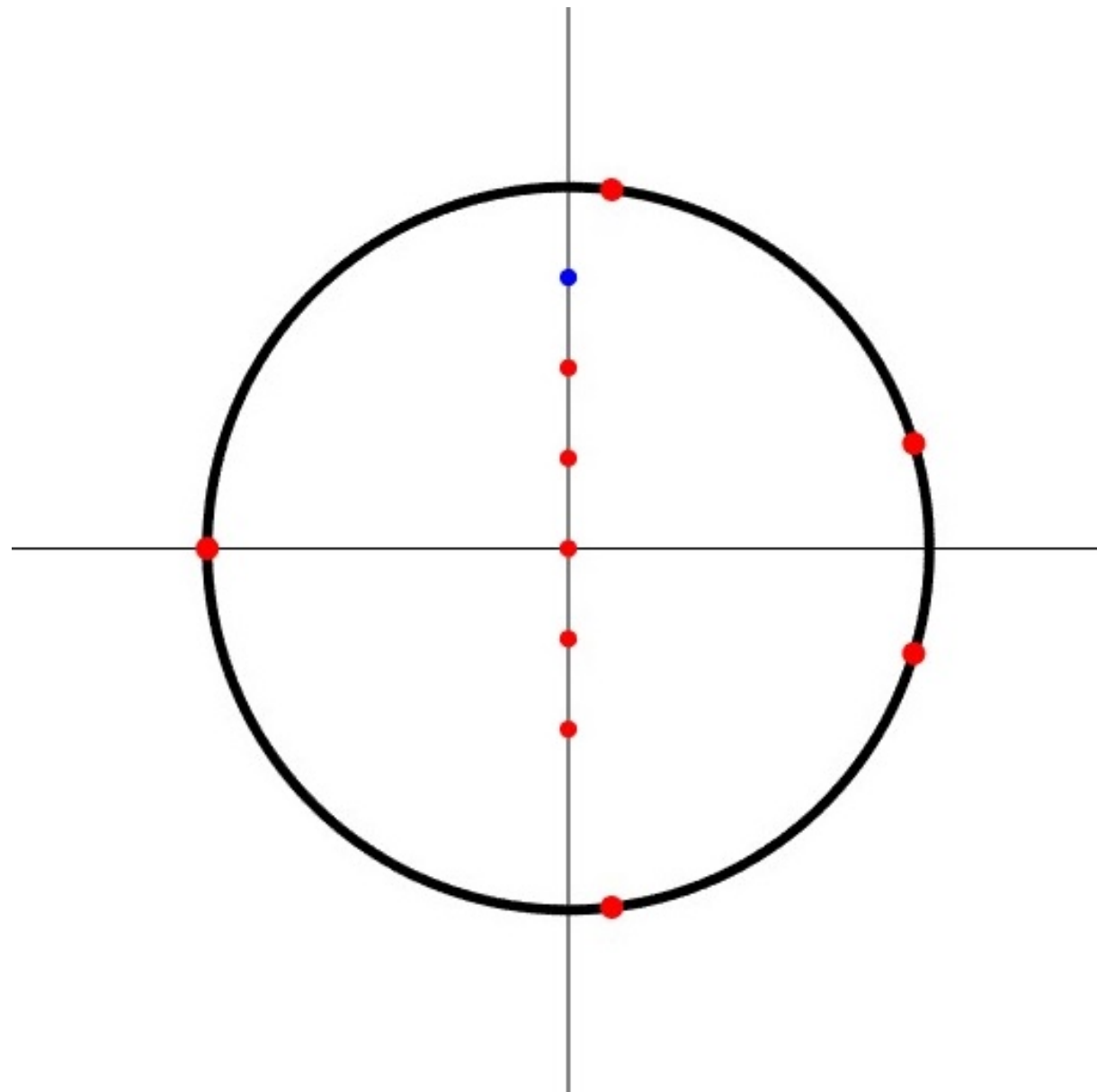


... and you get
 $n+1$ new ones!

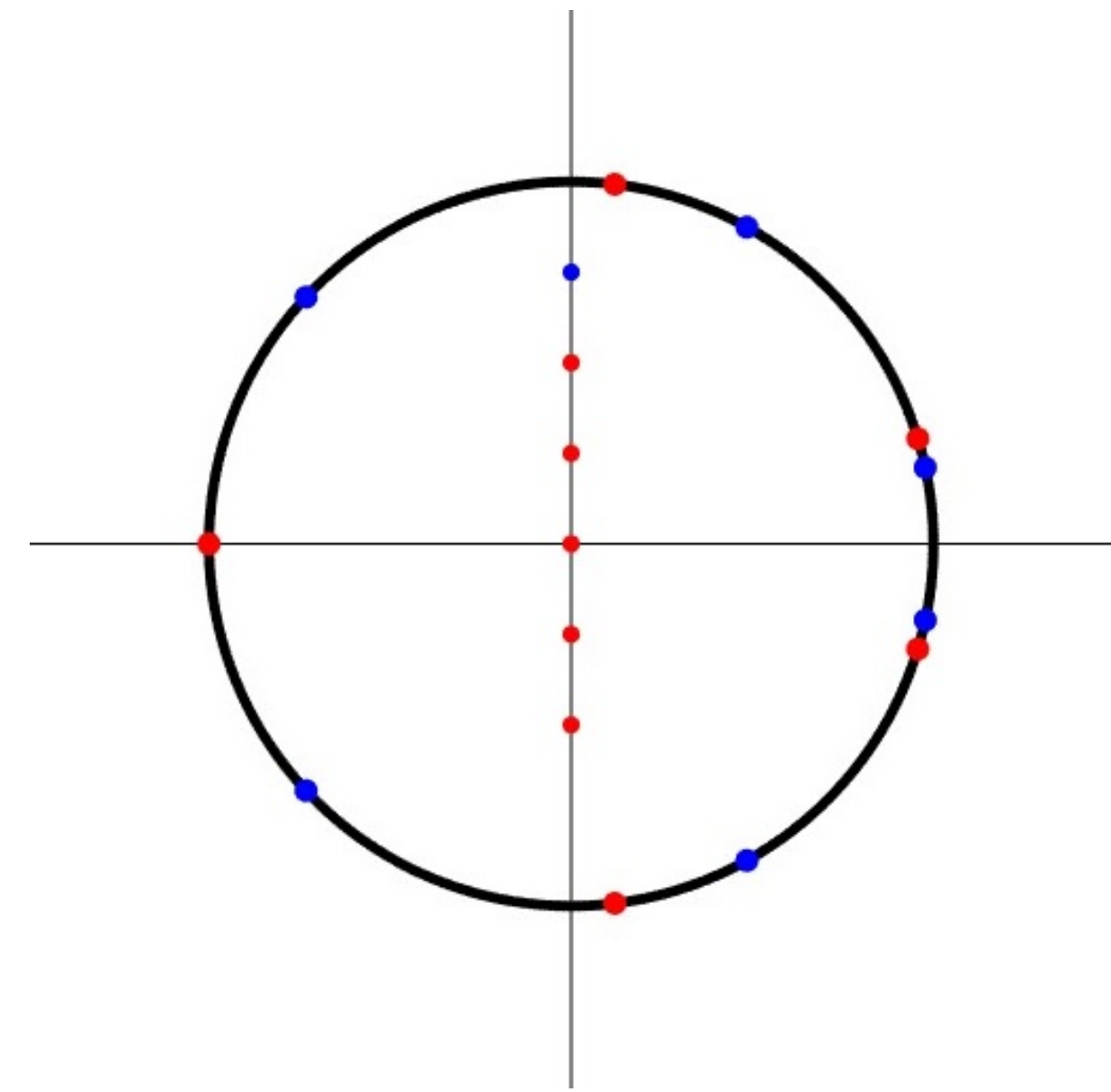


... A brave attempt, but with a rational ...

... simply add **one** (single) point ...



... and you get
n+1 new ones!



... extra tricks we can play ... (Laplace) **reparametrization!**

$$f(s_j) = \int_0^\infty e^{-ts} F(t) dt = t_0 \int_0^\infty e^{-tt_0 s} F(tt_0) dt = t_0 \int_0^\infty e^{-t} e^{-t(t_0 s - 1)} F(tt_0) dt$$

... we can get quite **a number of values!** ... actually **all those for which the machinery works** ...

... (Laplace) **reparametrization!**

... In the end ... **DOES IT WORK?!**

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Not only we can possibly get quite a number of values ... we get a **CONSISTENCY ARGUMENT!**

$$f(s_j) \sim \sum_k e^{-t_k(t_0 s - 1)} F(t_k t_0) dt$$

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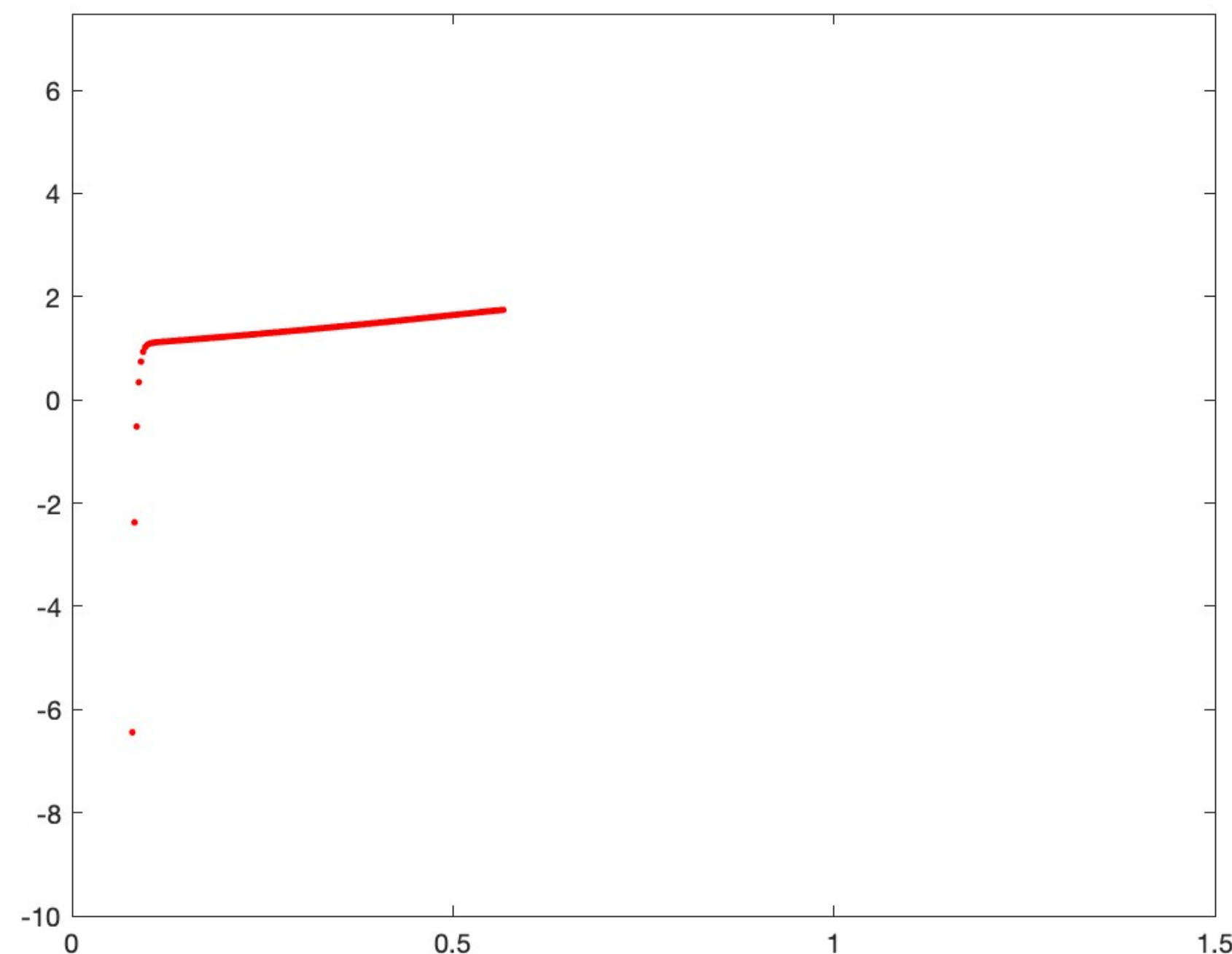
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$$f(s) = \frac{1}{s-1} \rightarrow F(t) = e^t$$

$$k = 2$$



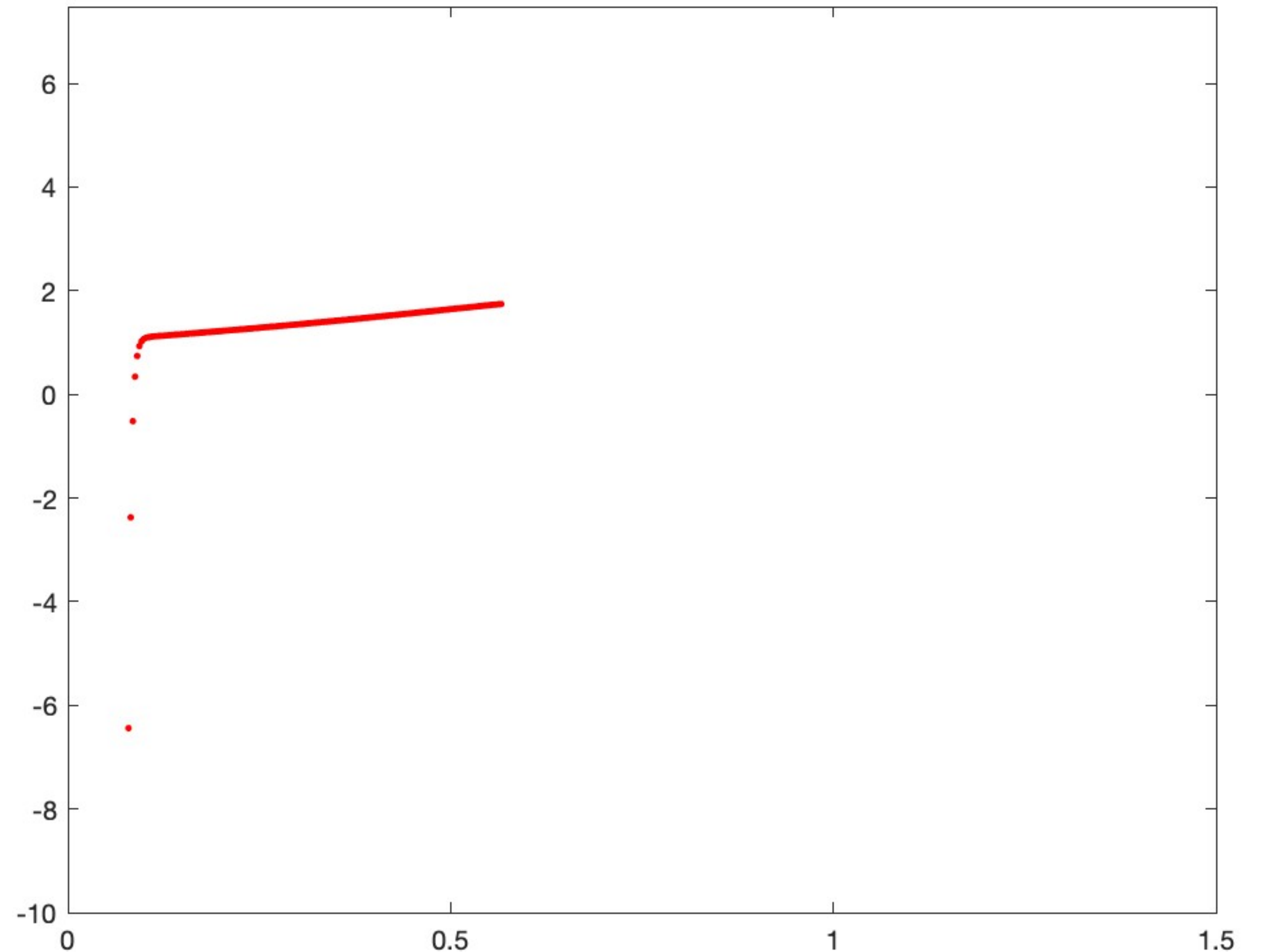
Some **CRAZY** points
... and ...
**a number falling on
a SMOOTH line**

... In the end ... DOES IT WORK?!

We said we get a **CONSISTENCY ARGUMENT** ...

$$f(s_j) \sim \sum_k e^{-t_k(t_0 s-1)} F(t_k t_0) dt$$
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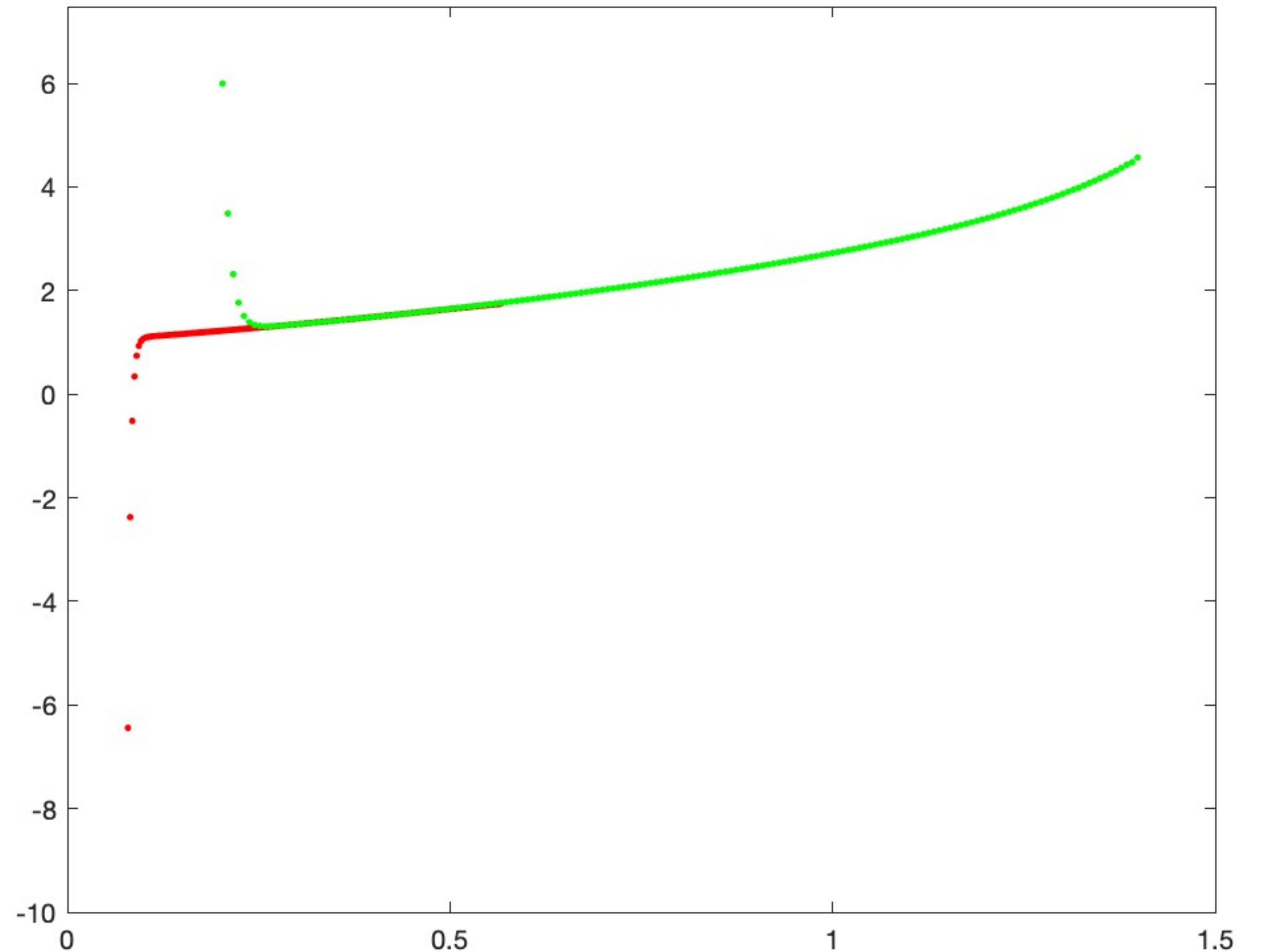


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$$k = 2, 3$$

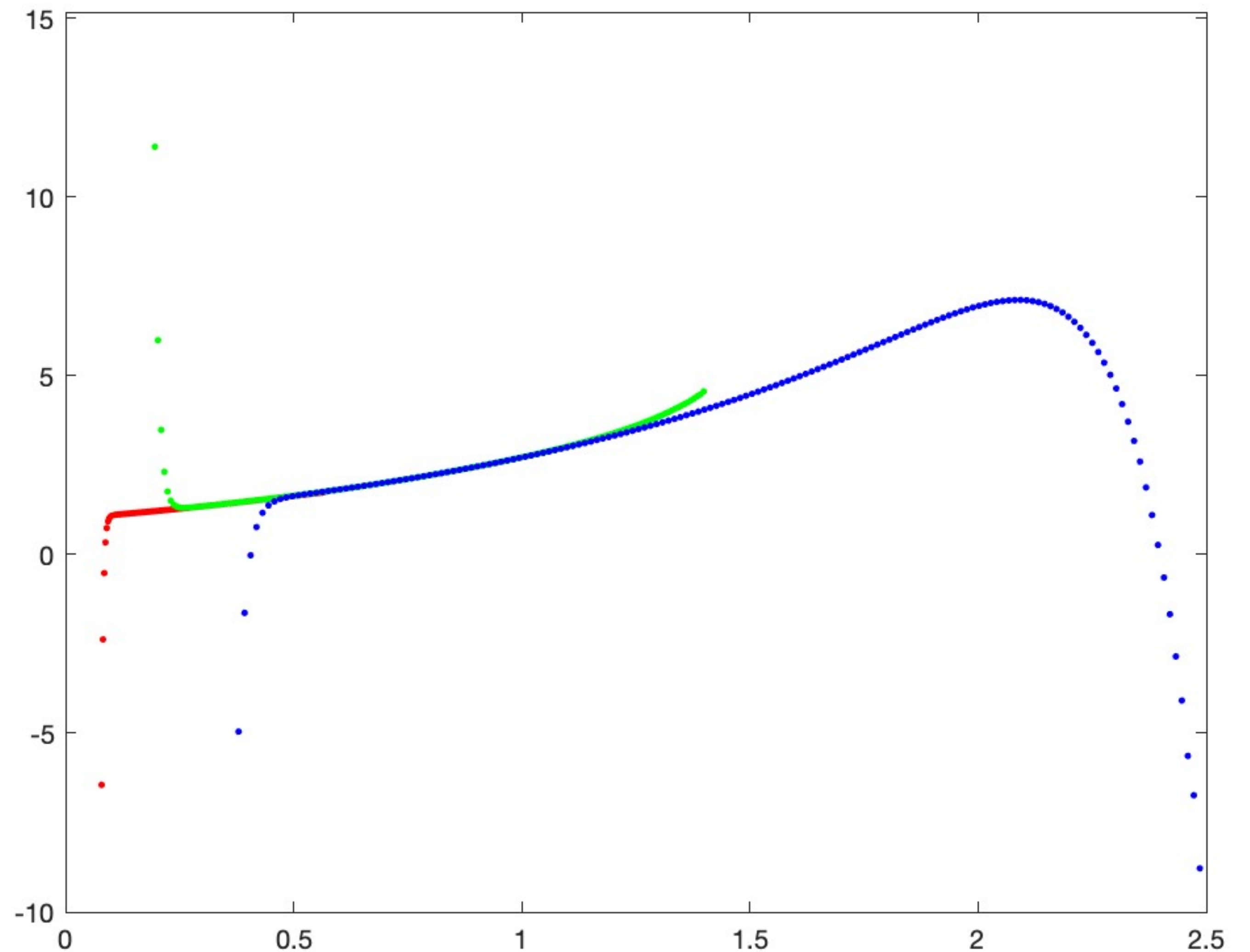


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$$k = 2, 3, 4$$



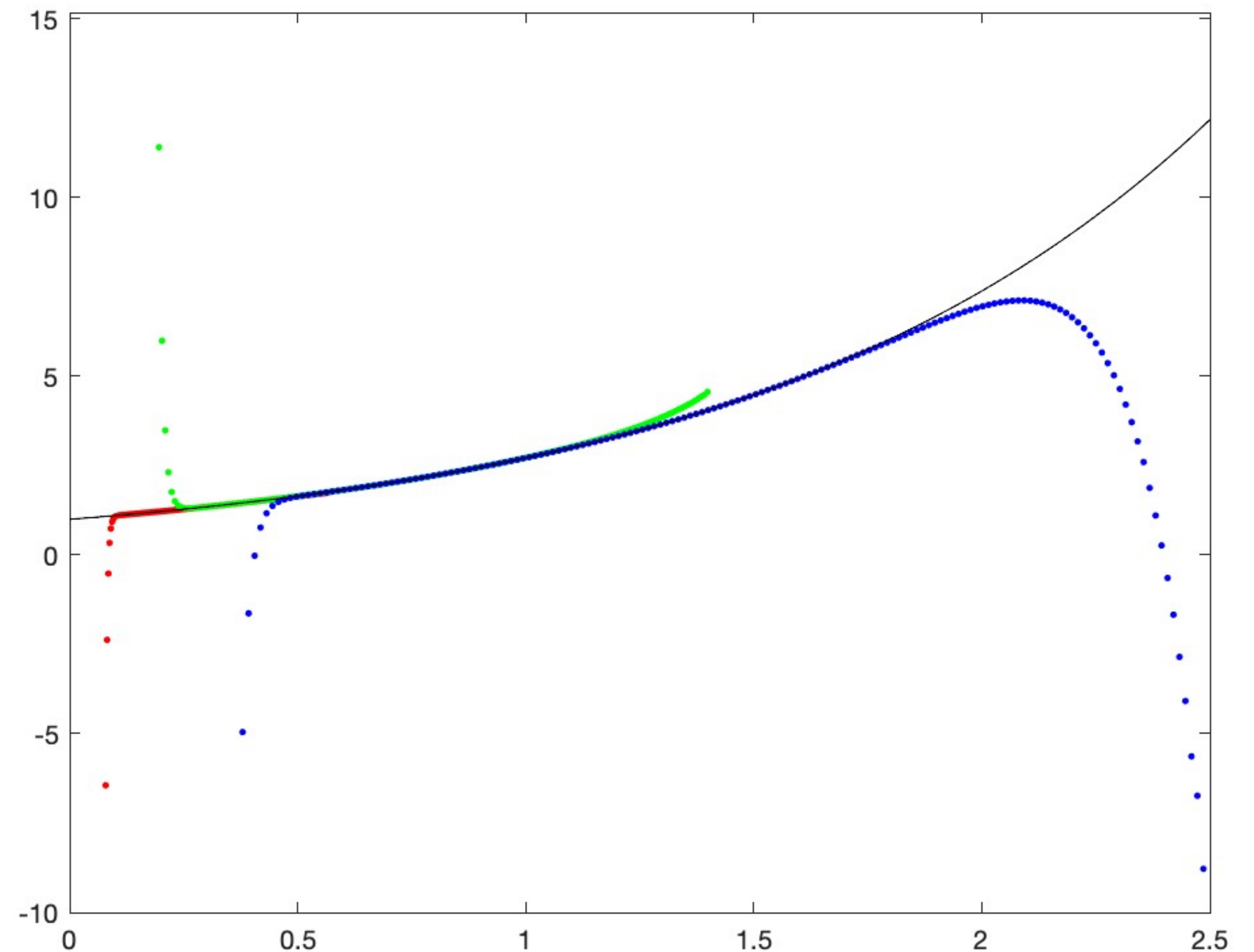
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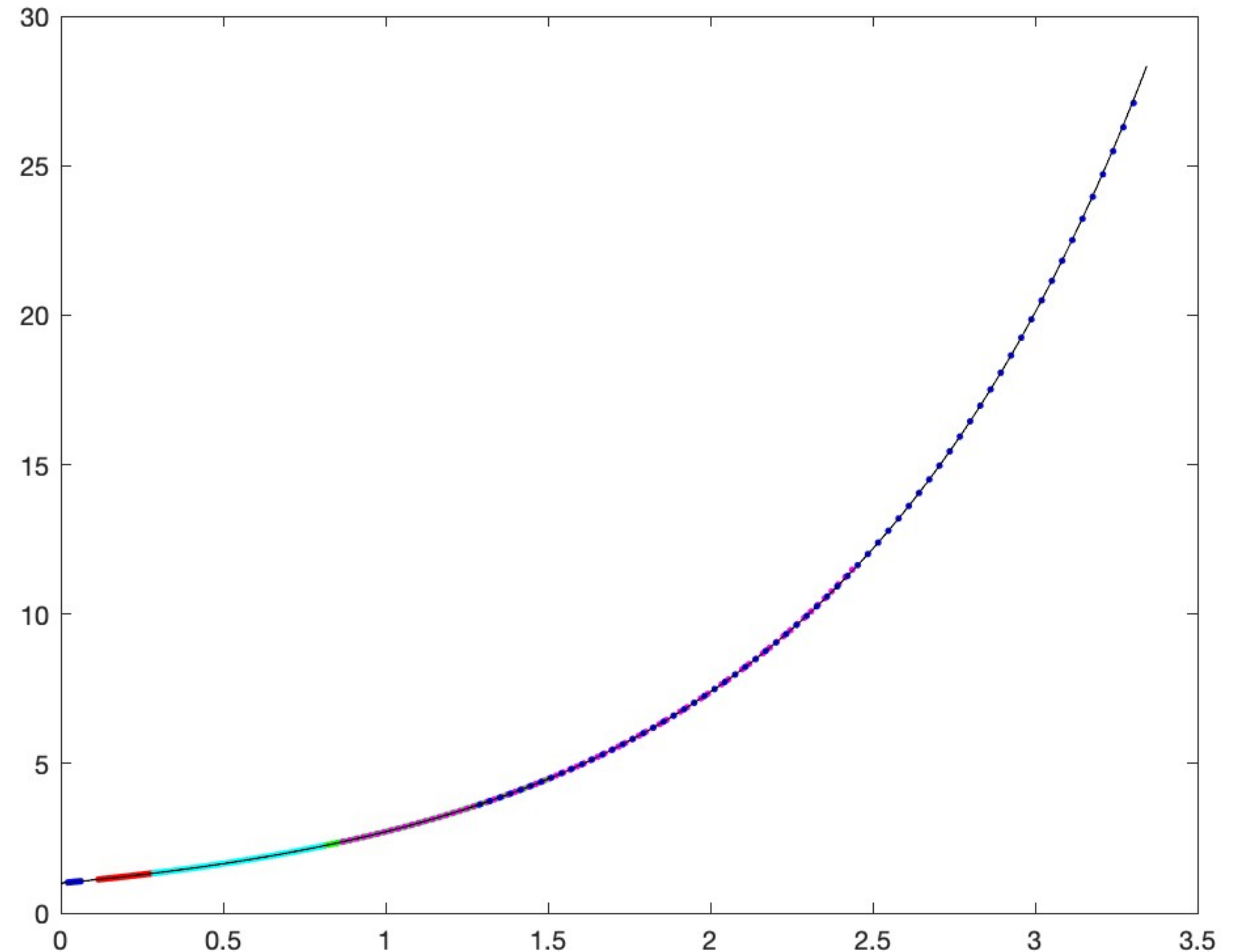
The black line is the **EXPONENTIAL** ...!



... In the end ... DOES IT WORK?!

... It is fair to say... IT WORKS!

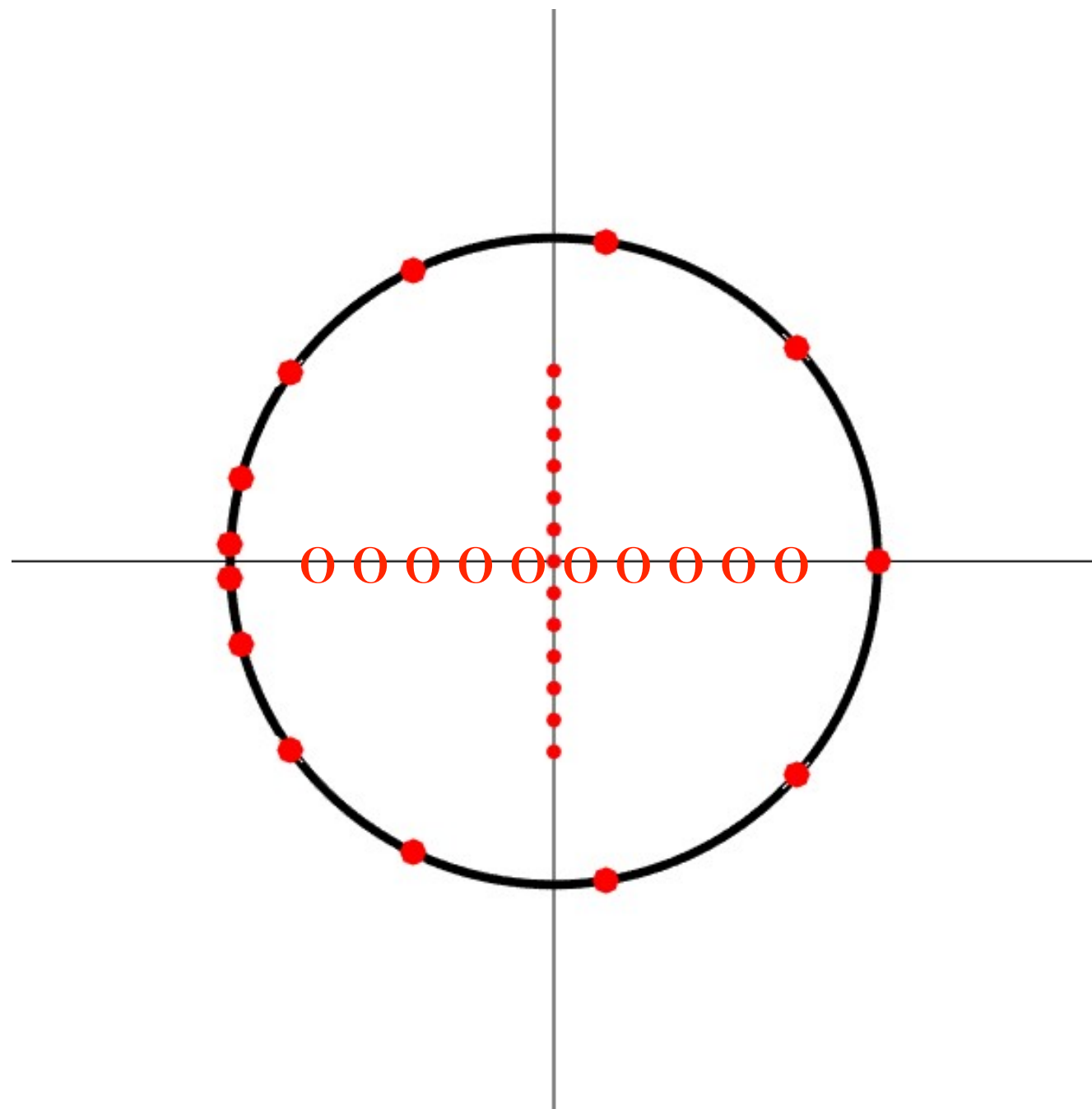
... provided we look for smooth overlaps ...!
(... of course, provided they show up ...)



... but what about the CAUCHY FORMULA?!

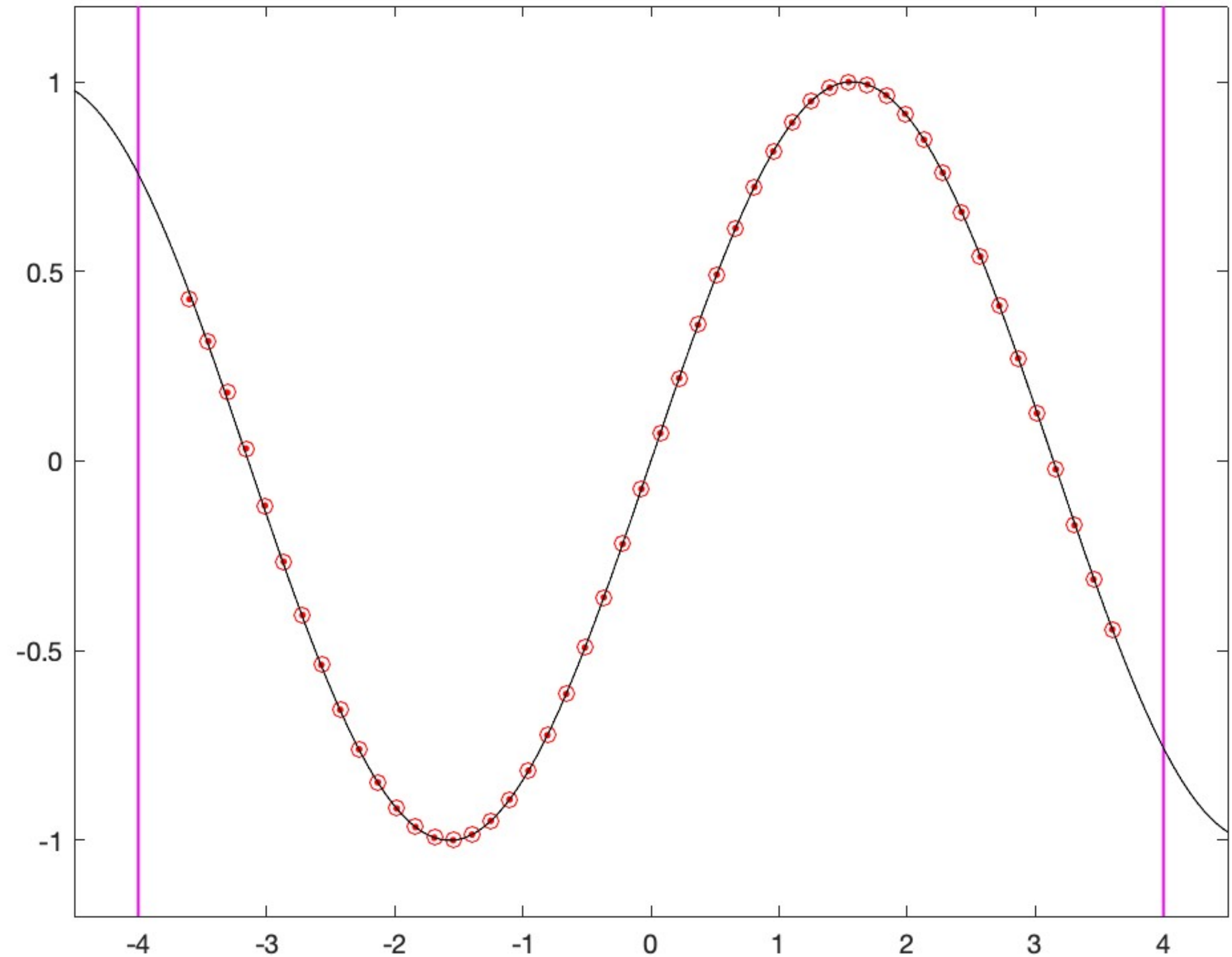
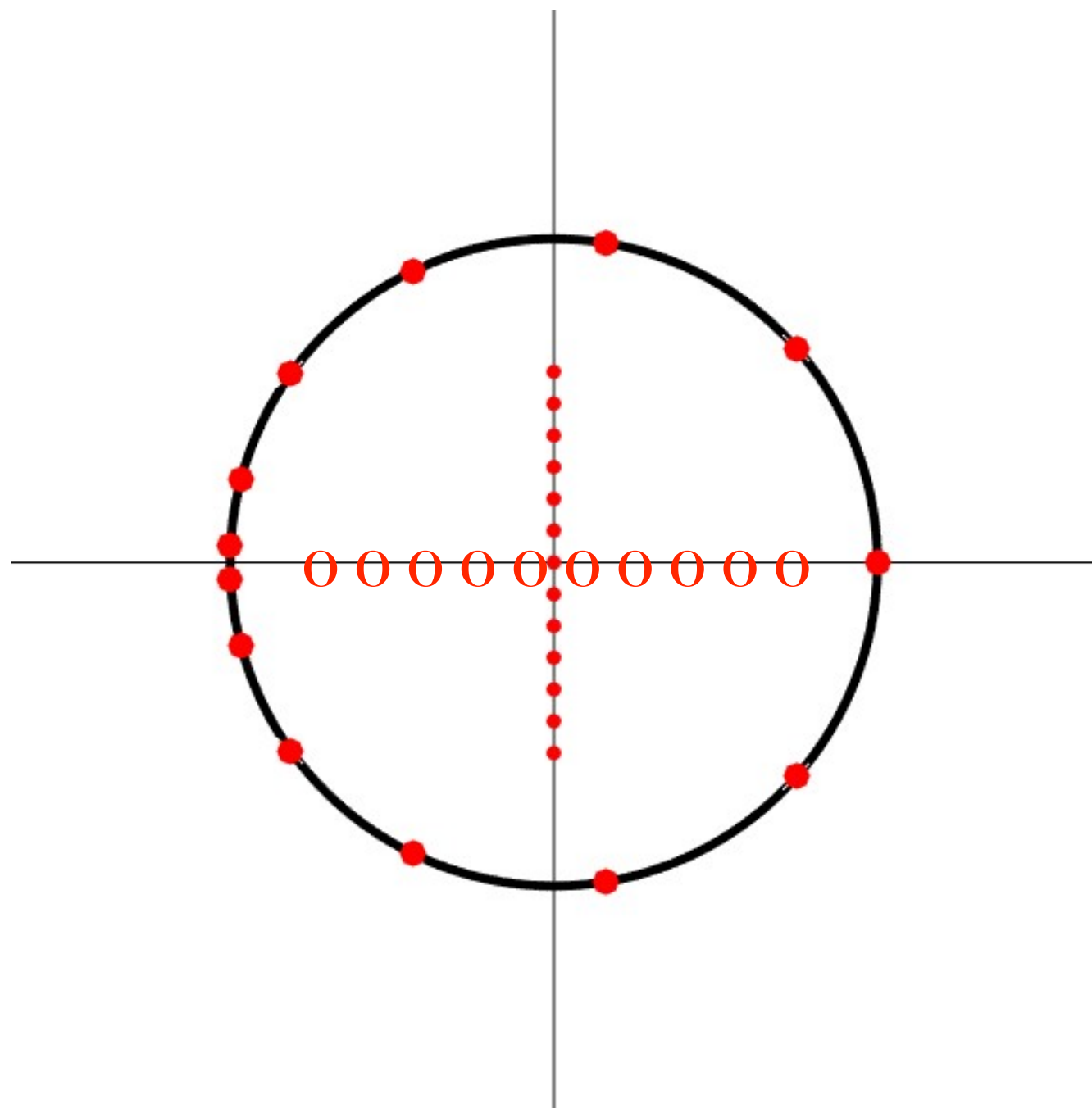
... but what about the CAUCHY FORMULA?!

This is the cartoon to remember ...



... but what about the **CAUCHY FORMULA?!**

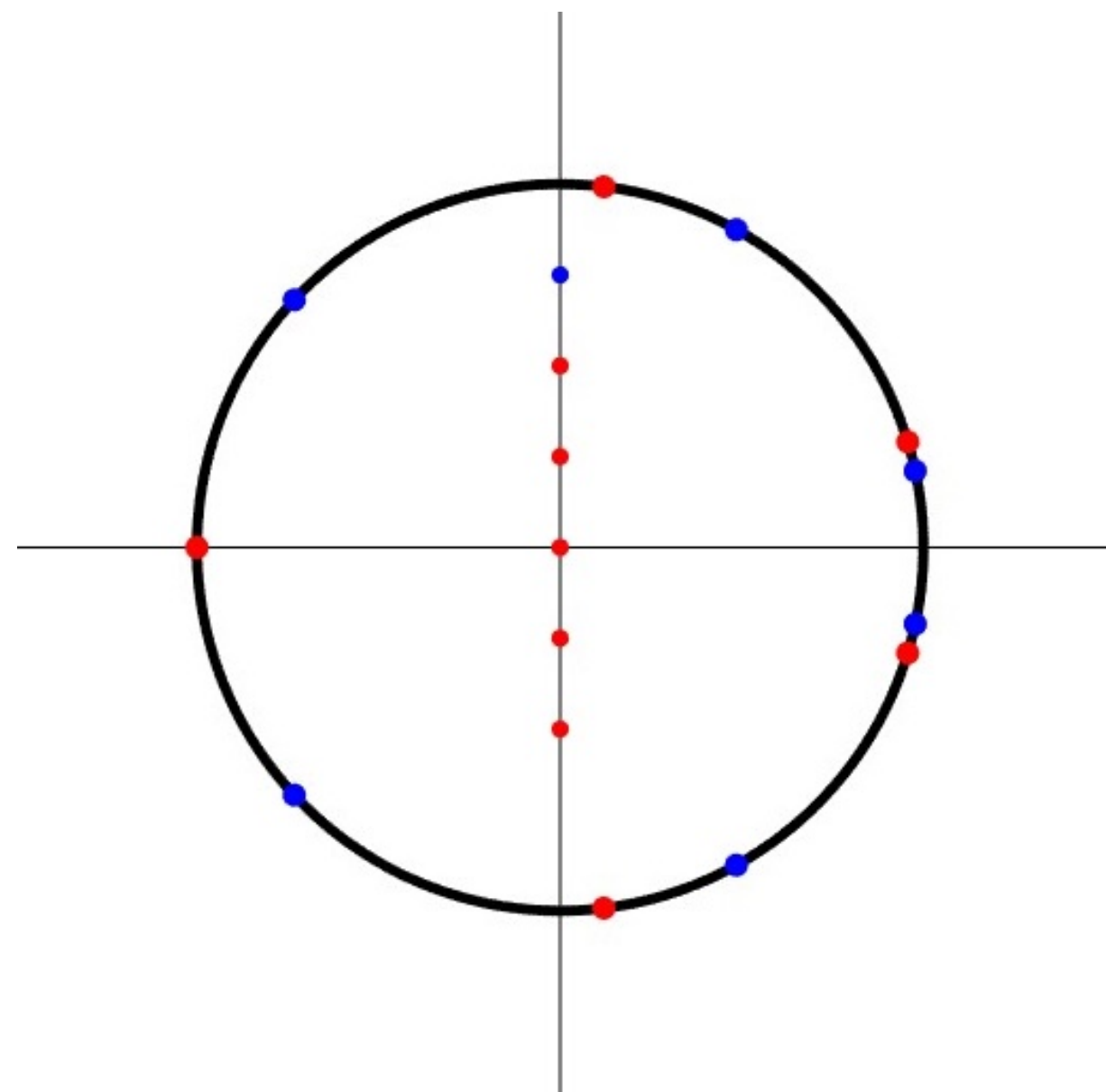
Here we go!... computing the sin function on the real axis
knowing values on the imaginary axis!



... but what about the CAUCHY FORMULA?! Can we trust this (apparently) good results?

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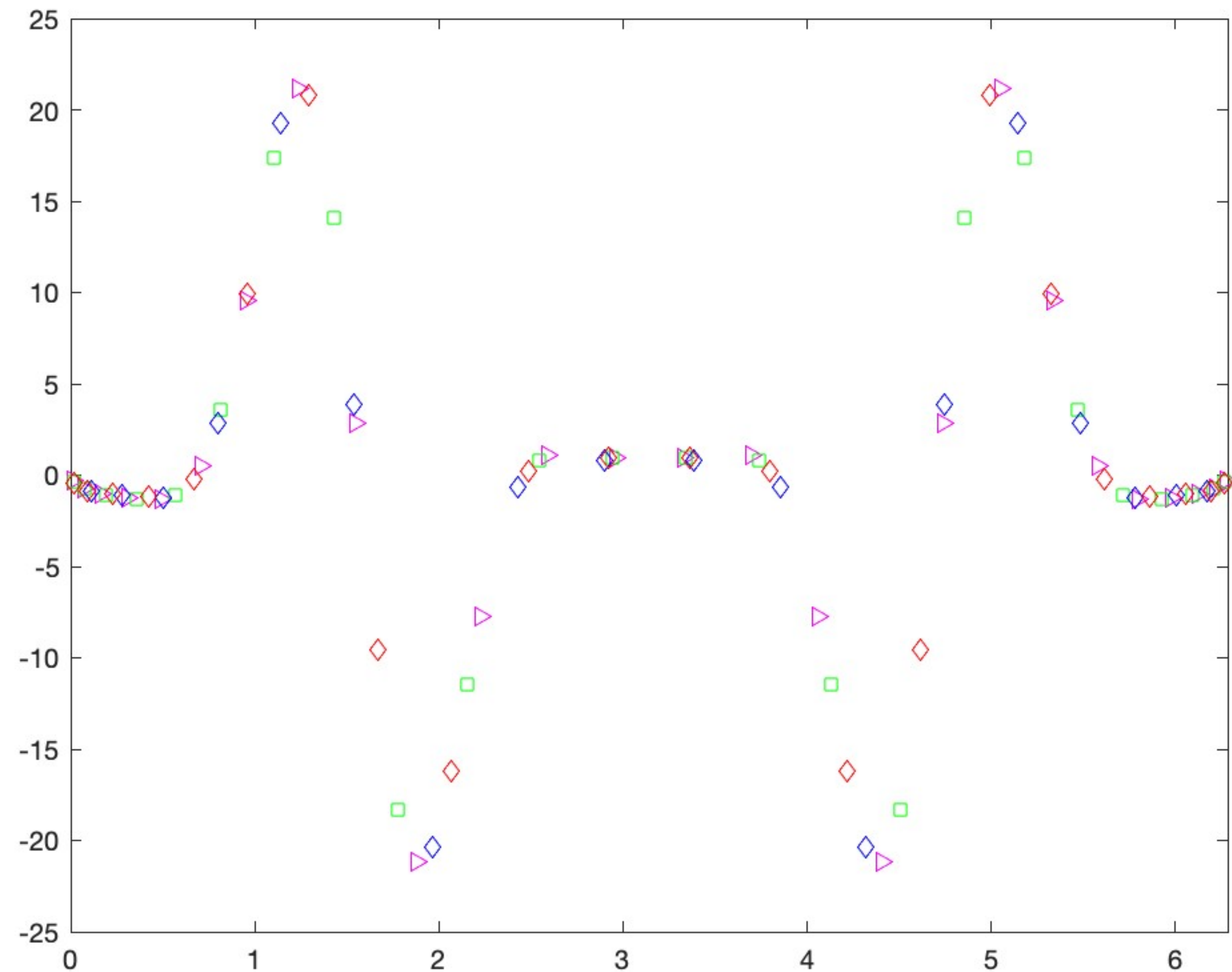
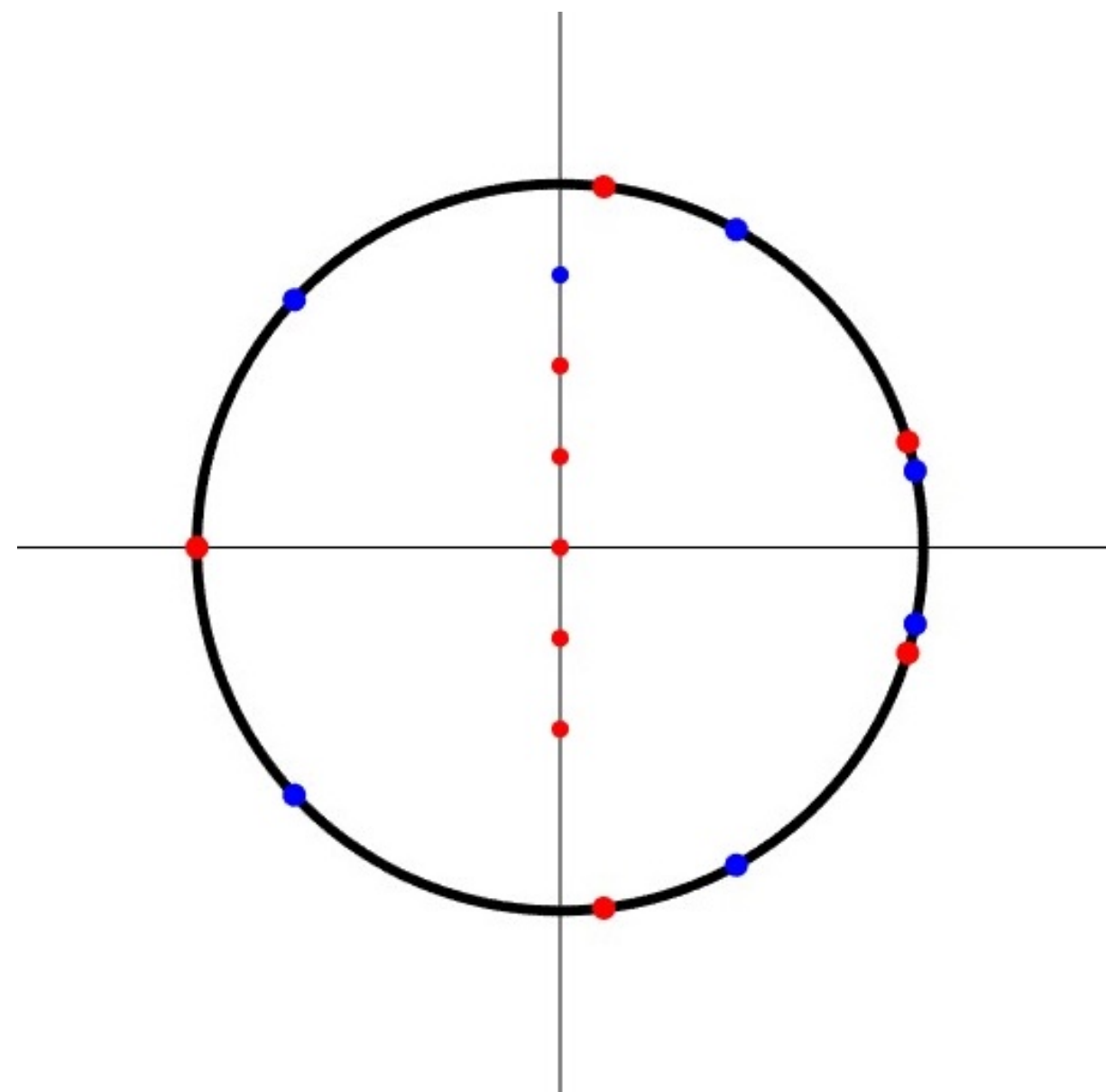
Now this is the cartoon to remember ...



... but what about the **CAUCHY FORMULA?! Can we trust this (apparently) good results?**

Looking at the **solution of the inverse problem**, we get a **smooth curve** showing up!

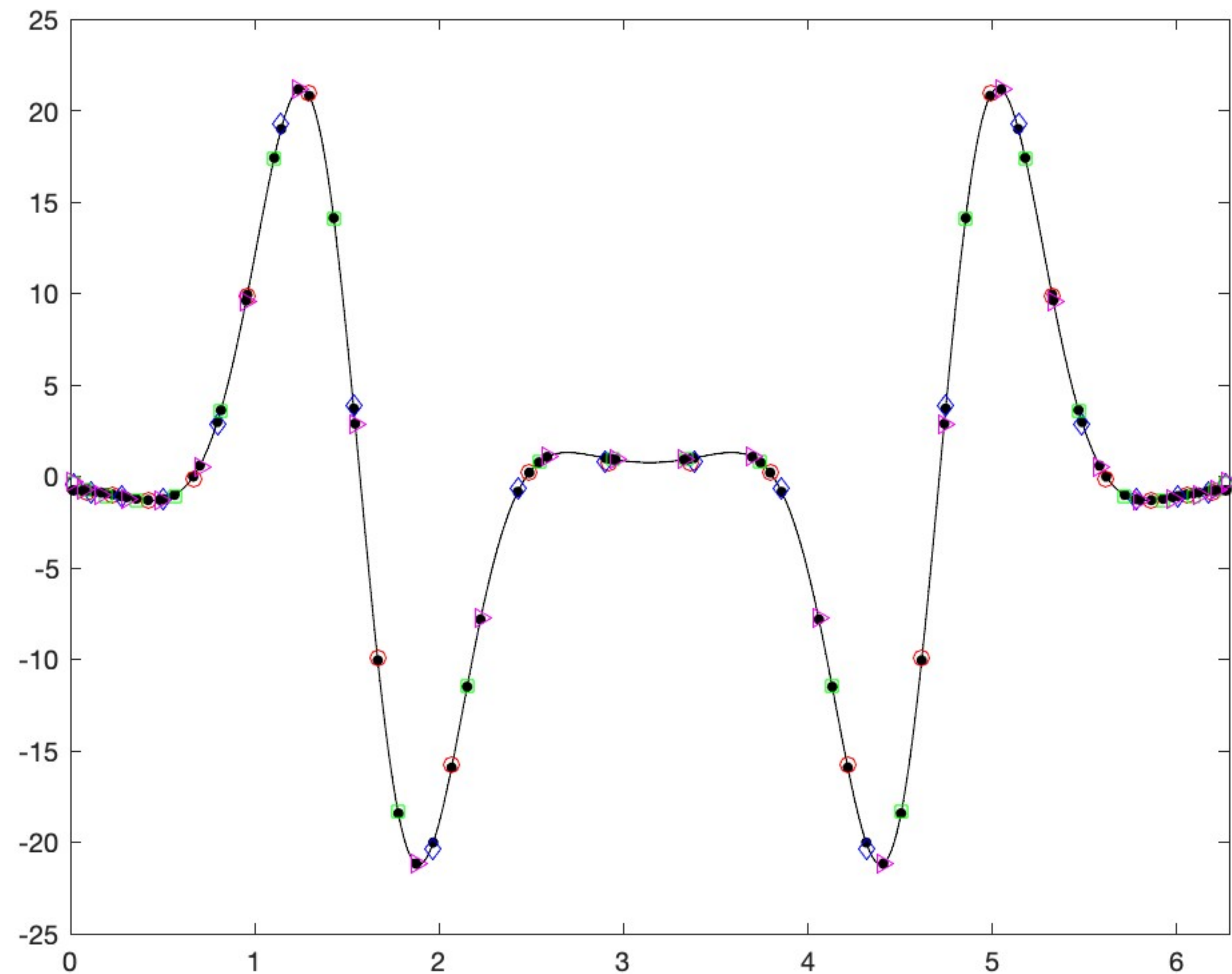
(This is the real part of the sin function at the quadrature points)

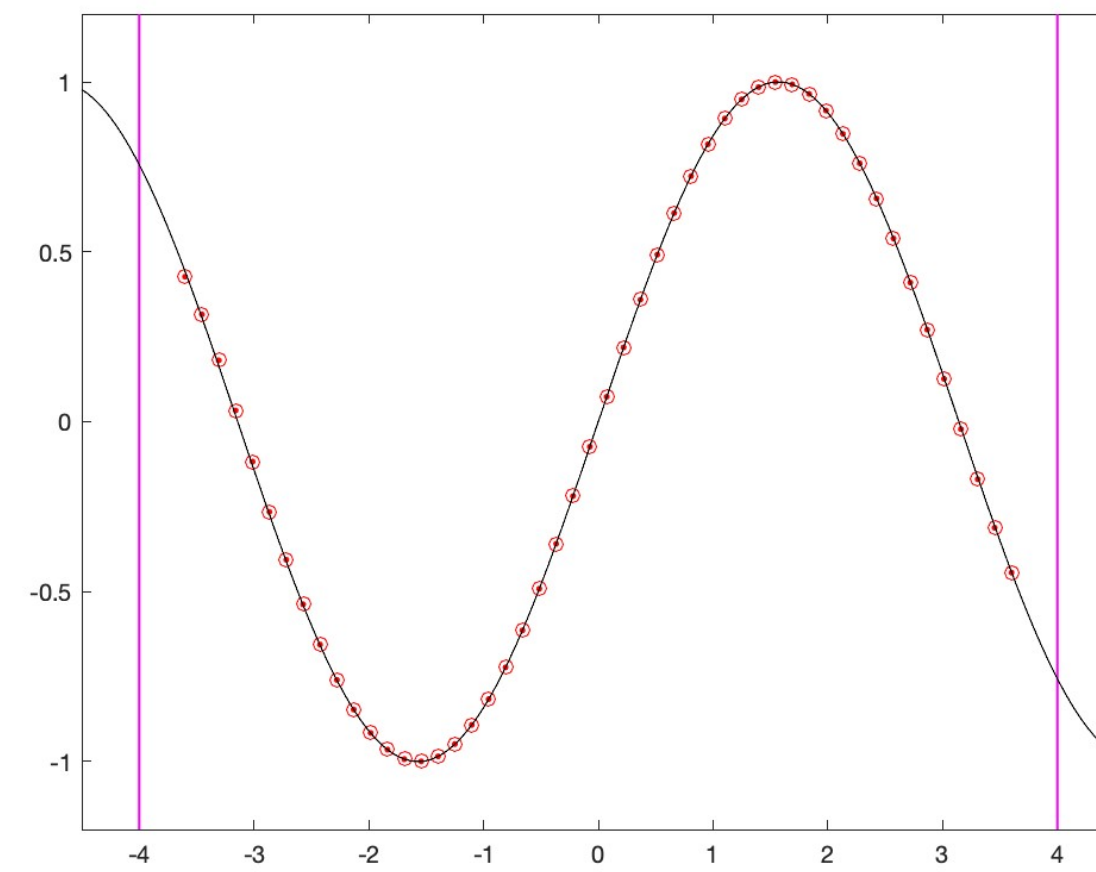
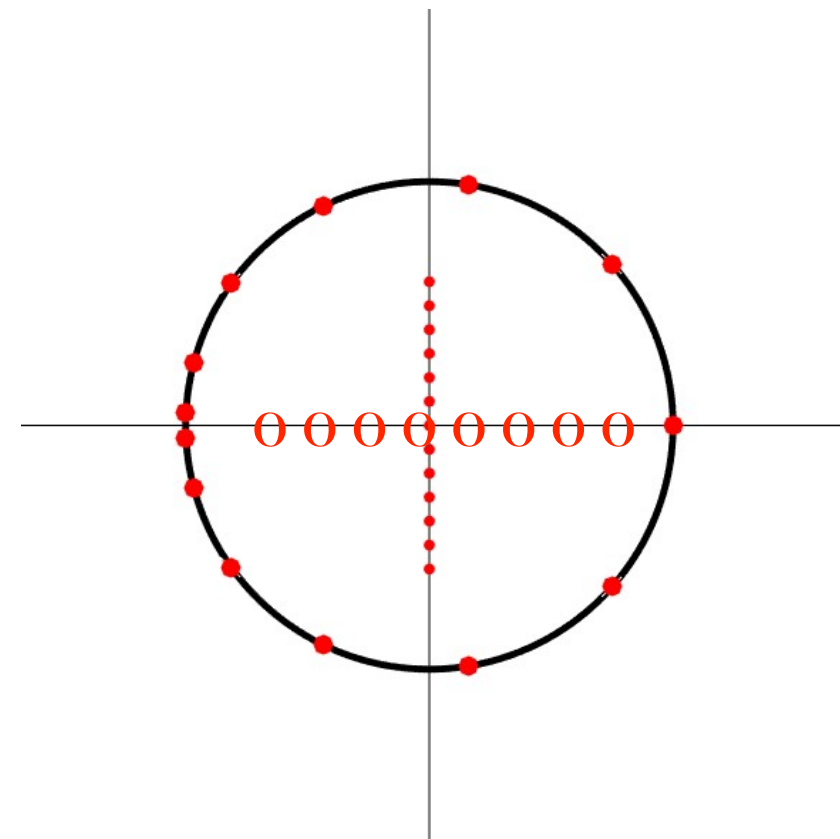


... but what about the **CAUCHY FORMULA?! Can we trust this (apparently) good results? YES!**

(This is the real part of the sin function at the quadrature points)

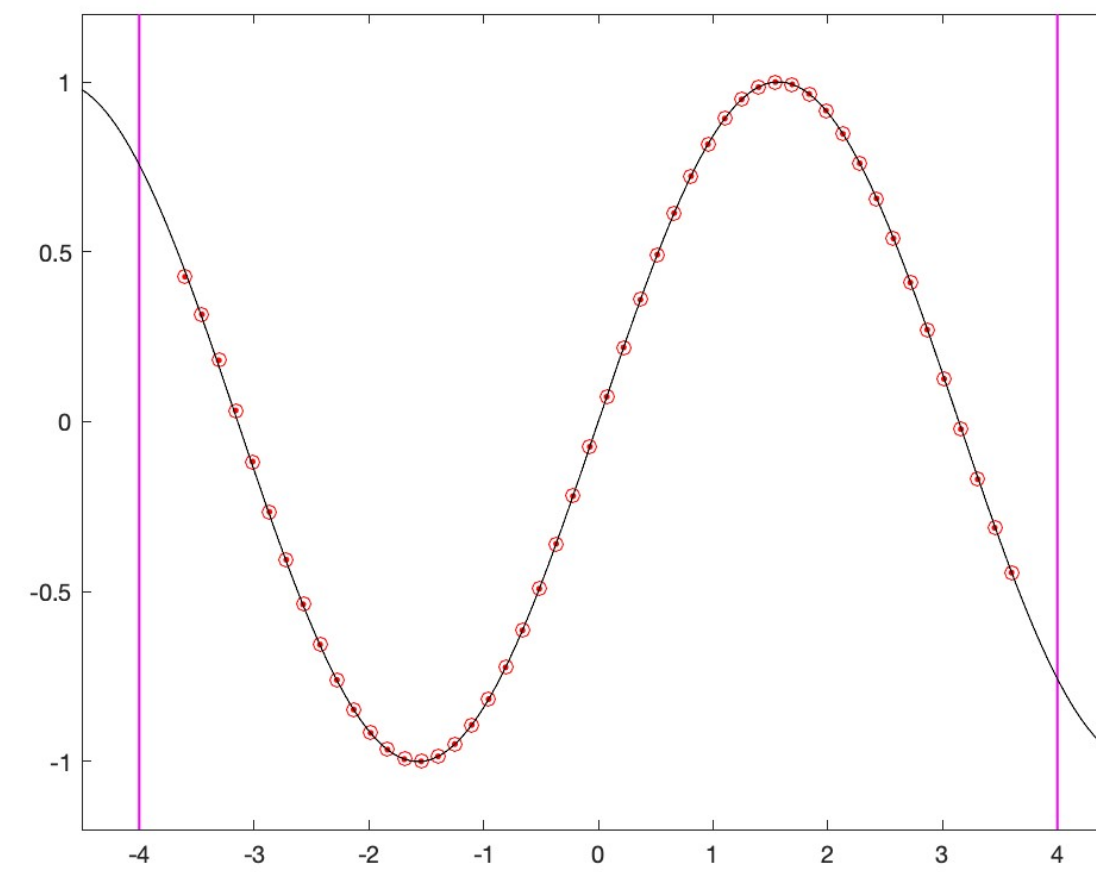
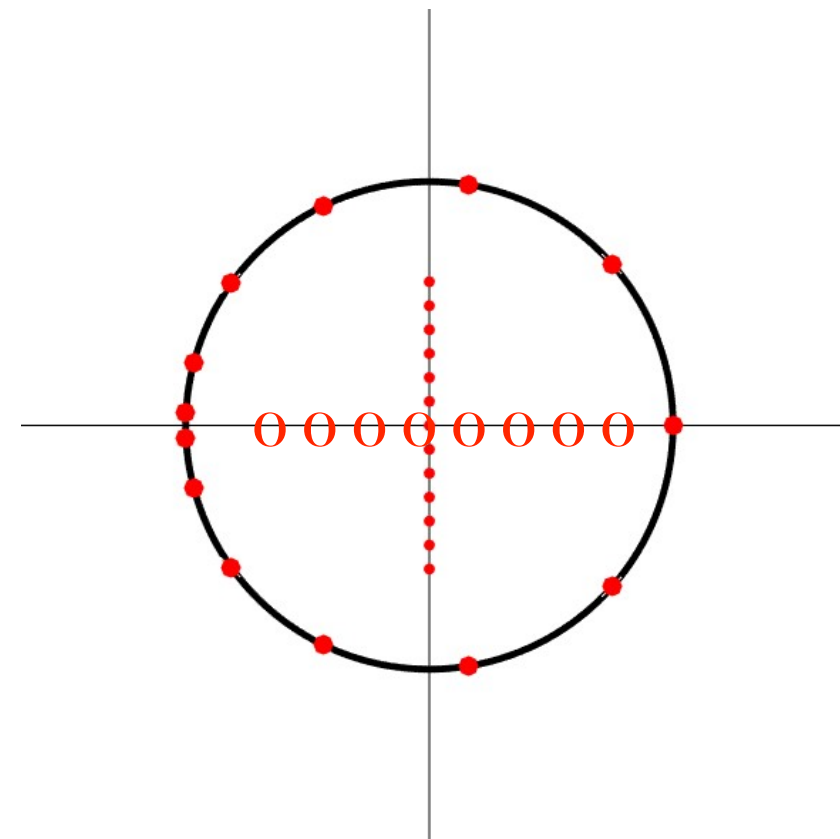
It is indeed what we
should have found ...





I have not yet told you the entire story ...

Remember... computing the sin function on the real axis
knowing values on the imaginary axis!

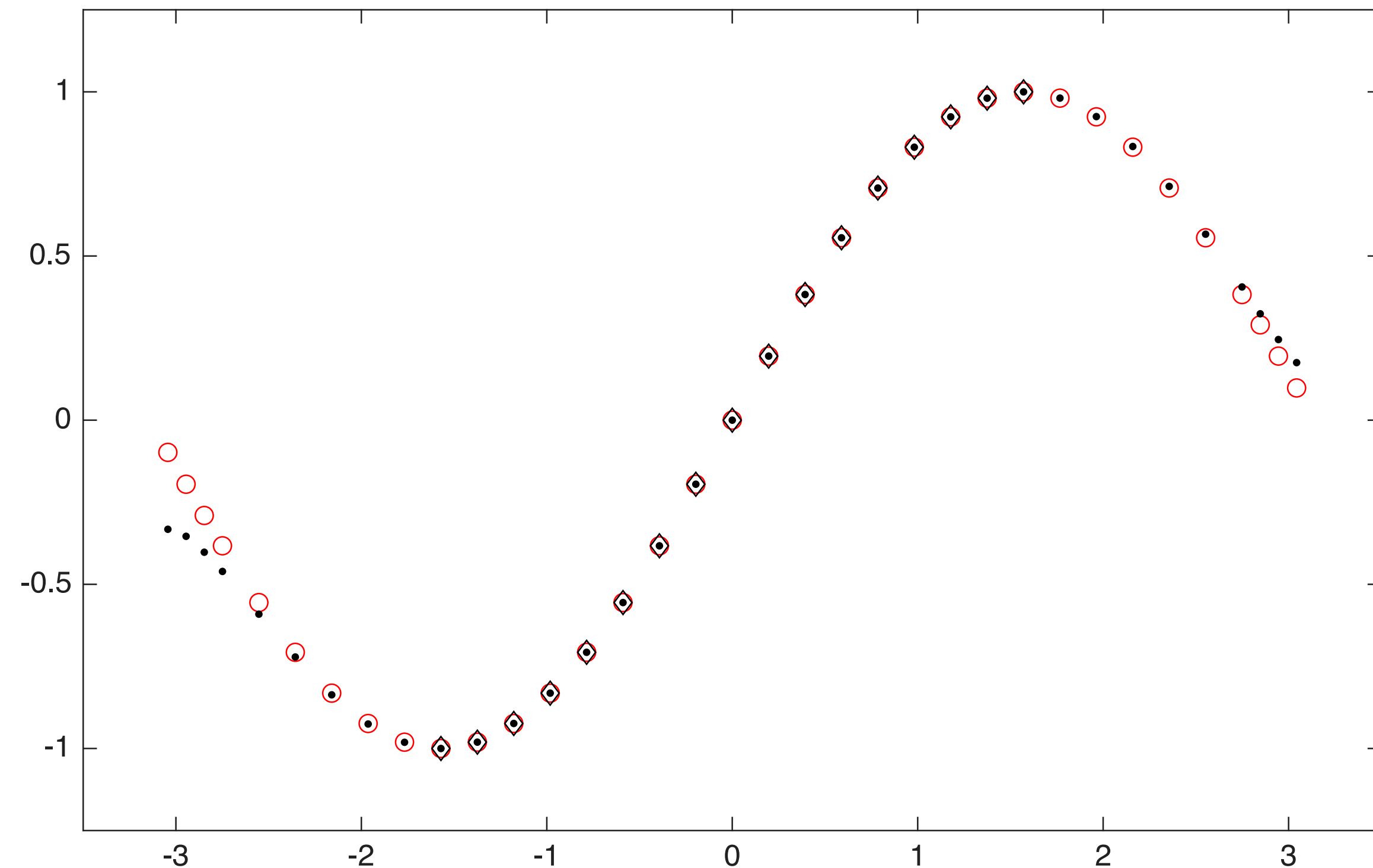


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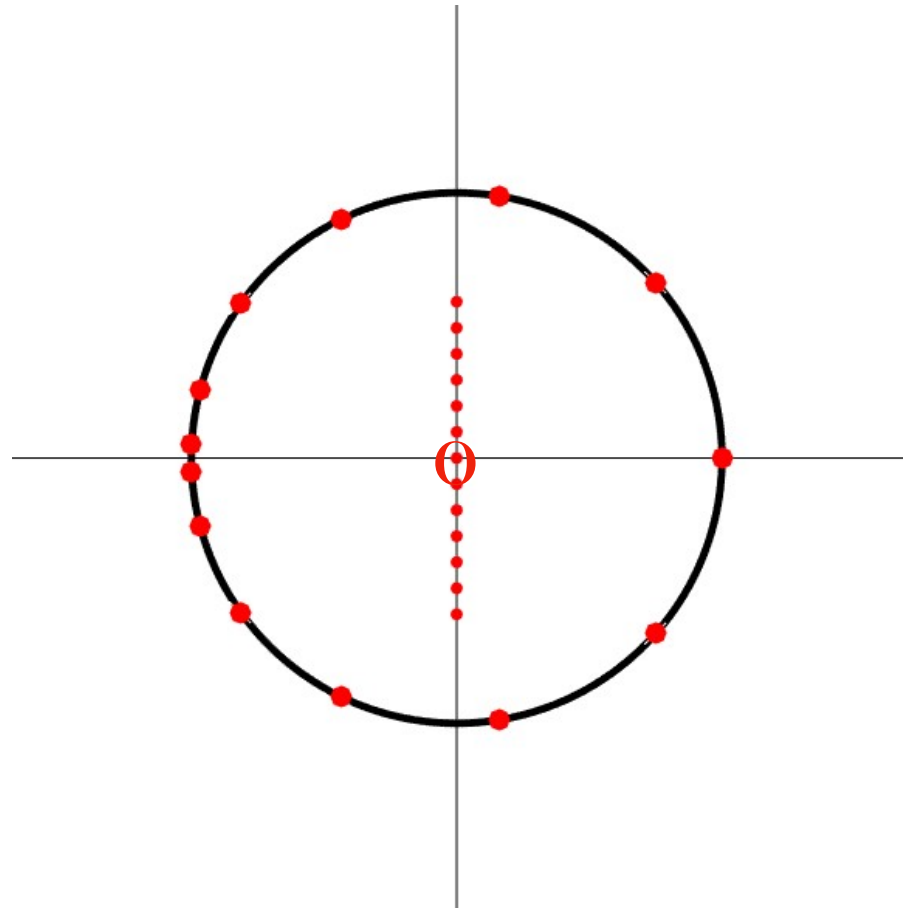
... and so ... first of all ...

LET'S LIVE WITH LESS INFORMATION ...



Remember!

once you solved for the function at the quadrature points ...



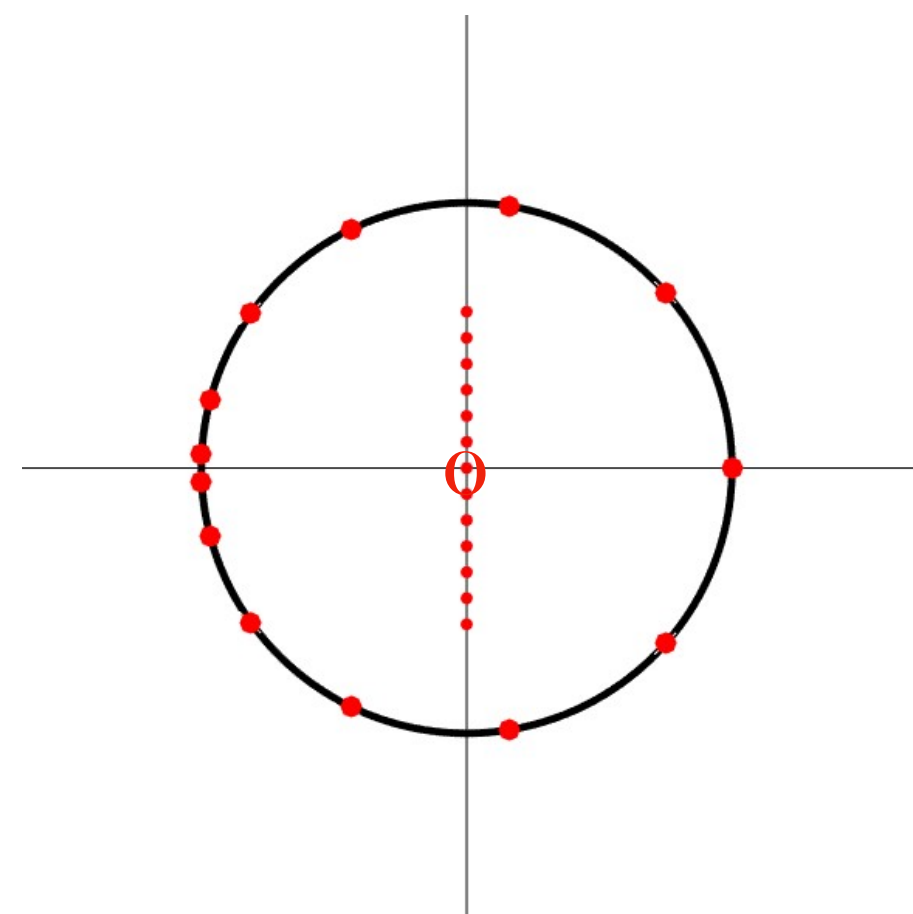
$$f(z_0) = \frac{1}{2\pi} \int_0^{2\pi} \frac{f(R e^{i\theta}) R e^{i\theta}}{R e^{i\theta} - z_0} d\theta \simeq \frac{1}{2\pi} \sum_{k=1}^n w_k \frac{f(R e^{i\theta_k}) R e^{i\theta_k}}{R e^{i\theta_k} - z_0}$$

$$y_i = \frac{1}{2\pi} \sum_{k=1}^n w_k \frac{R e^{i\theta_k}}{R e^{i\theta_k} - z_i} \hat{f}_k, \quad i = 1, 2, \dots, n$$

$$A \mathbf{x} = \mathbf{b}$$

... you have derivatives (e.g. in zero) for free!

$$f^{(n)}(z_0) = \frac{n!}{2\pi i} \oint_C \frac{f(z)}{(z - z_0)^{n+1}} dz = \frac{n!}{2\pi i} \int_0^{2\pi} \frac{f(R \exp(i\theta)) R \exp(i\theta)}{(R \exp i\theta - z_0)^{n+1}} d\theta$$



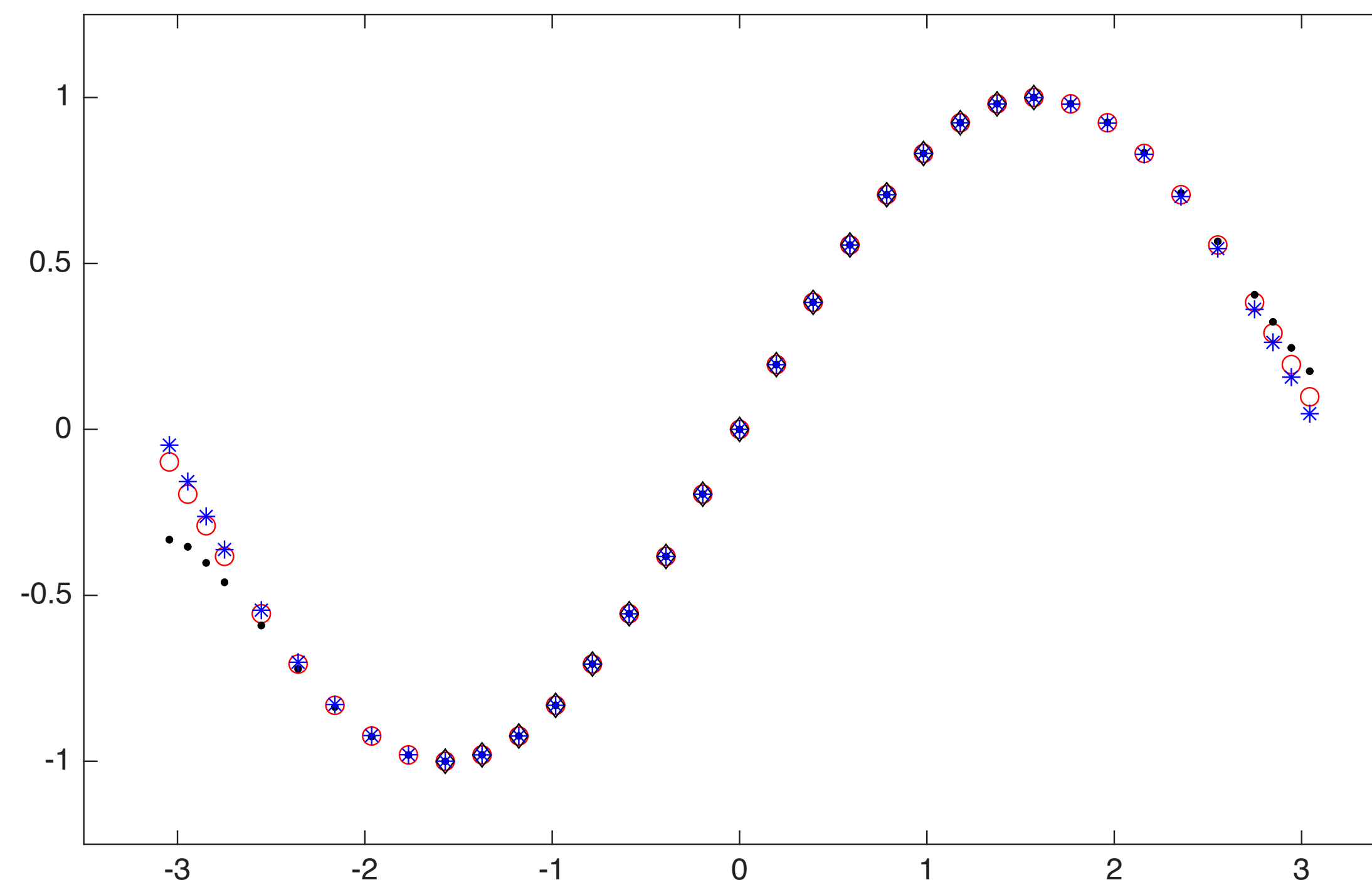
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... and you LIVE WITH LESS INFORMATION
by simply **COMPUTING TAYLOR SERIES!**

computing the sin function on the real axis

This does not come as a surprise ... the series
converges fast enough to live with only a
moderate number of terms ... and you can
also verify the convergence pattern!

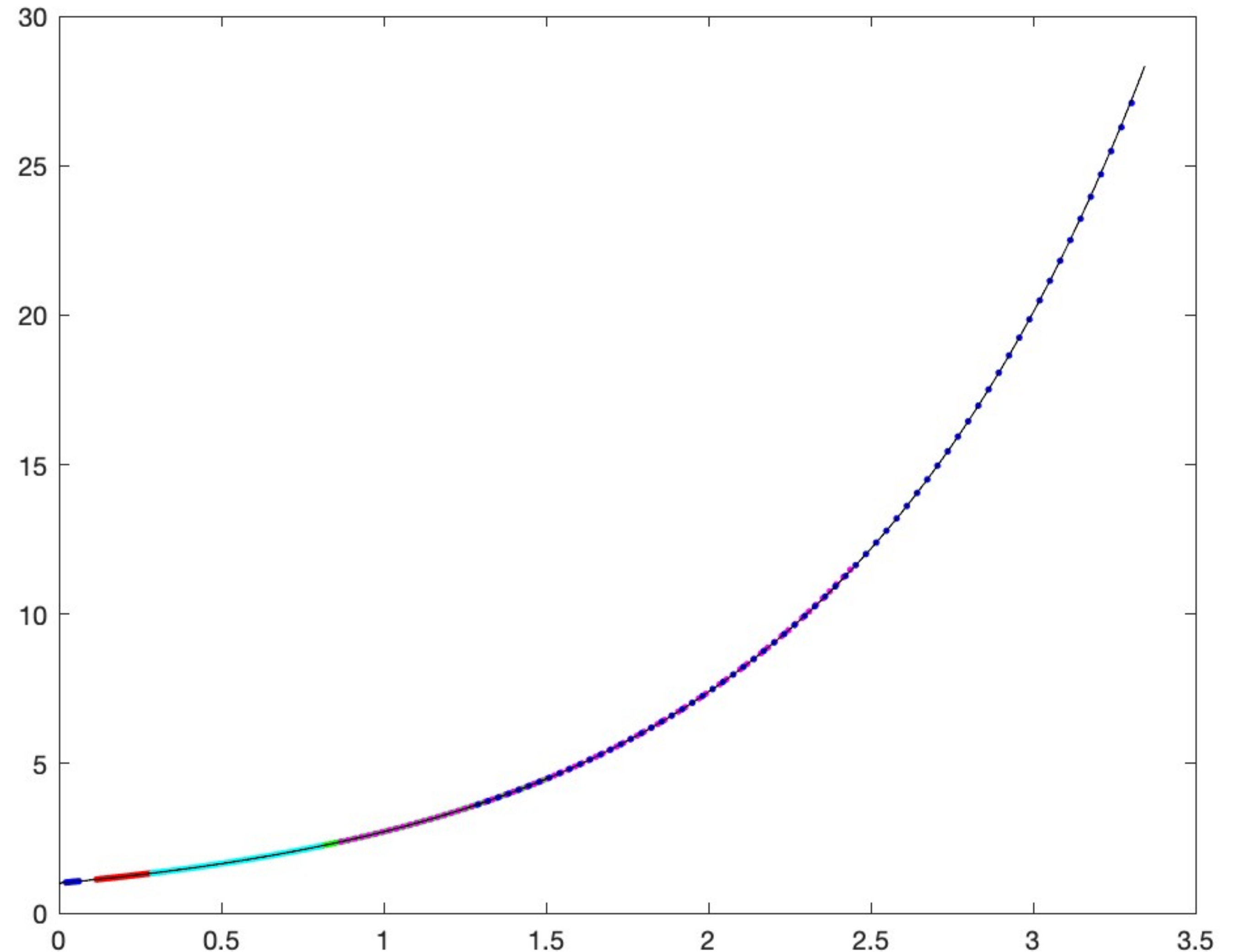


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Do you remember the **great success** ?
(Anti)Laplace Transformation

$$f(s_j) \sim \sum_k e^{-t_k(t_0 s - 1)} F(t_k t_0) dt$$
$$f(s) = \frac{1}{s-1} \quad \rightarrow \quad F(t) = e^t$$

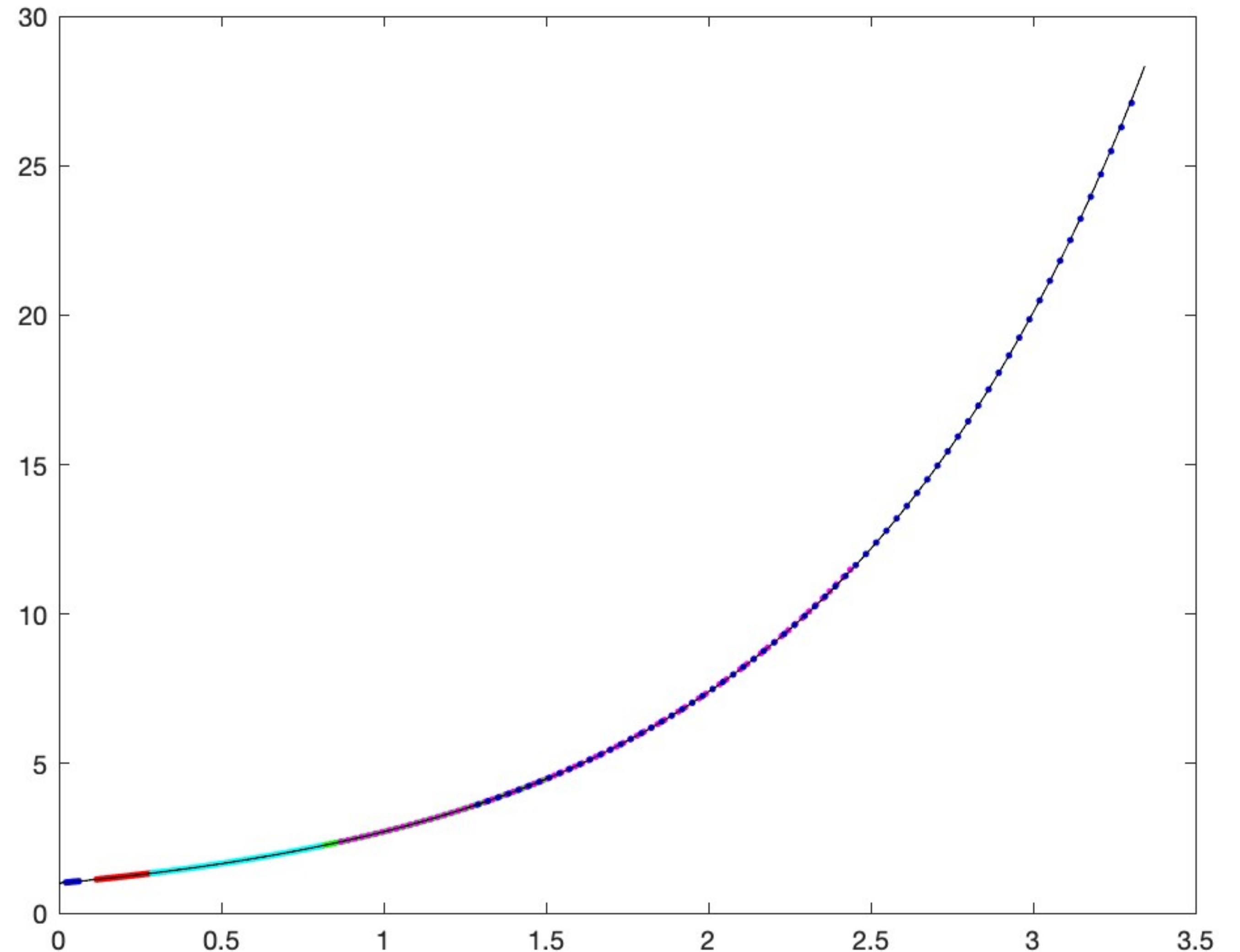


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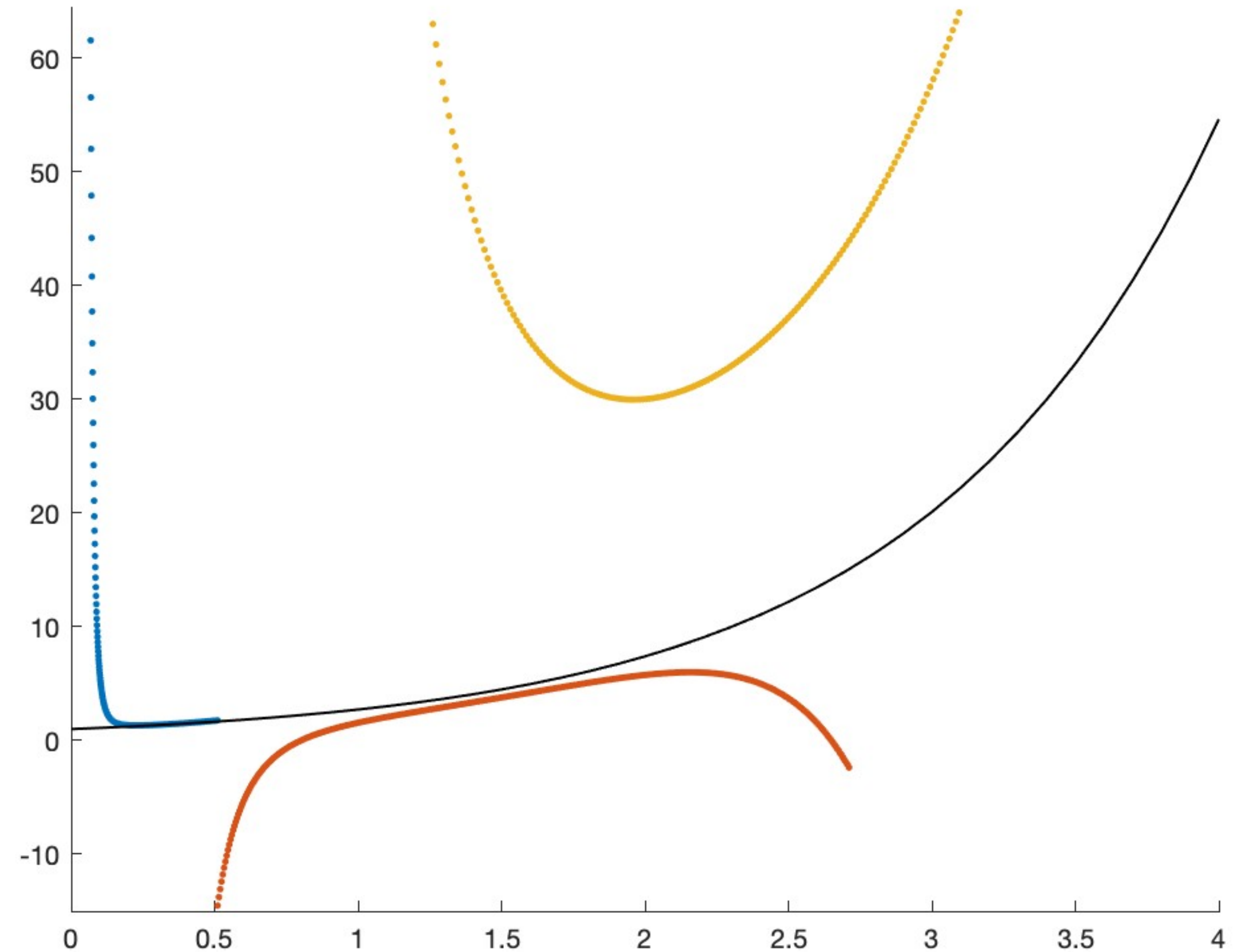
... so, what if **ERRORS** show up !?!



I have not yet told you the entire story ...

... so, what if **ERRORS** show up ???

Apparently, you **loose everything** !



Again, not a surprise ... when a linear system is **ill-conditioned**, tiny errors can result in a *solution* which is far away from the real one!

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We go for **TYCHONOV REGULARISATION**, i.e.

to solve $A\mathbf{x} = \mathbf{b}$

you minimise $||A\mathbf{x} - \mathbf{b}||^2 + ||\gamma\mathbf{x}||^2$

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... but then ... what is your choice for γ ?

to solve

$$A\mathbf{x} = \mathbf{b}$$

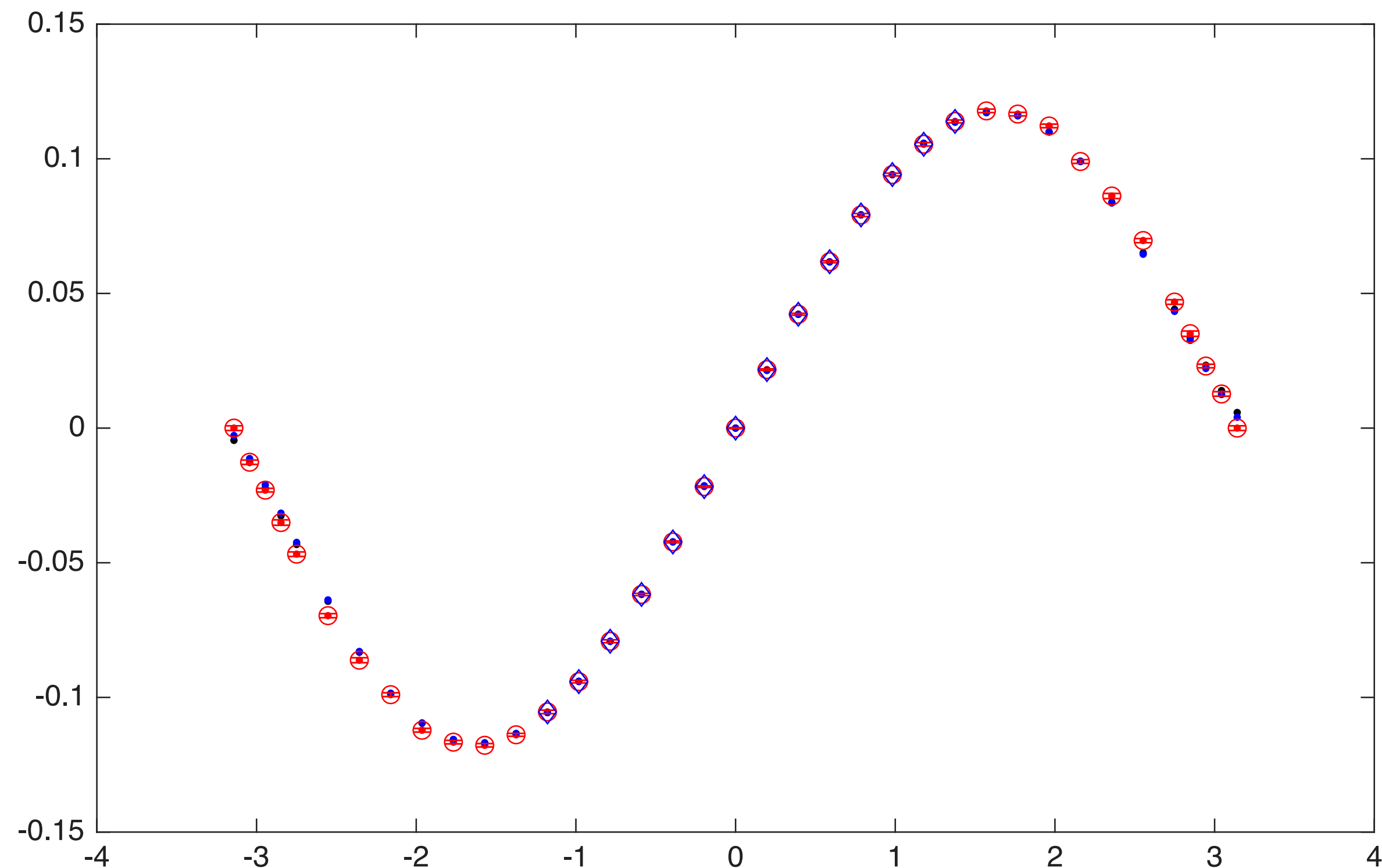
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... and we tune γ to best describe our input data

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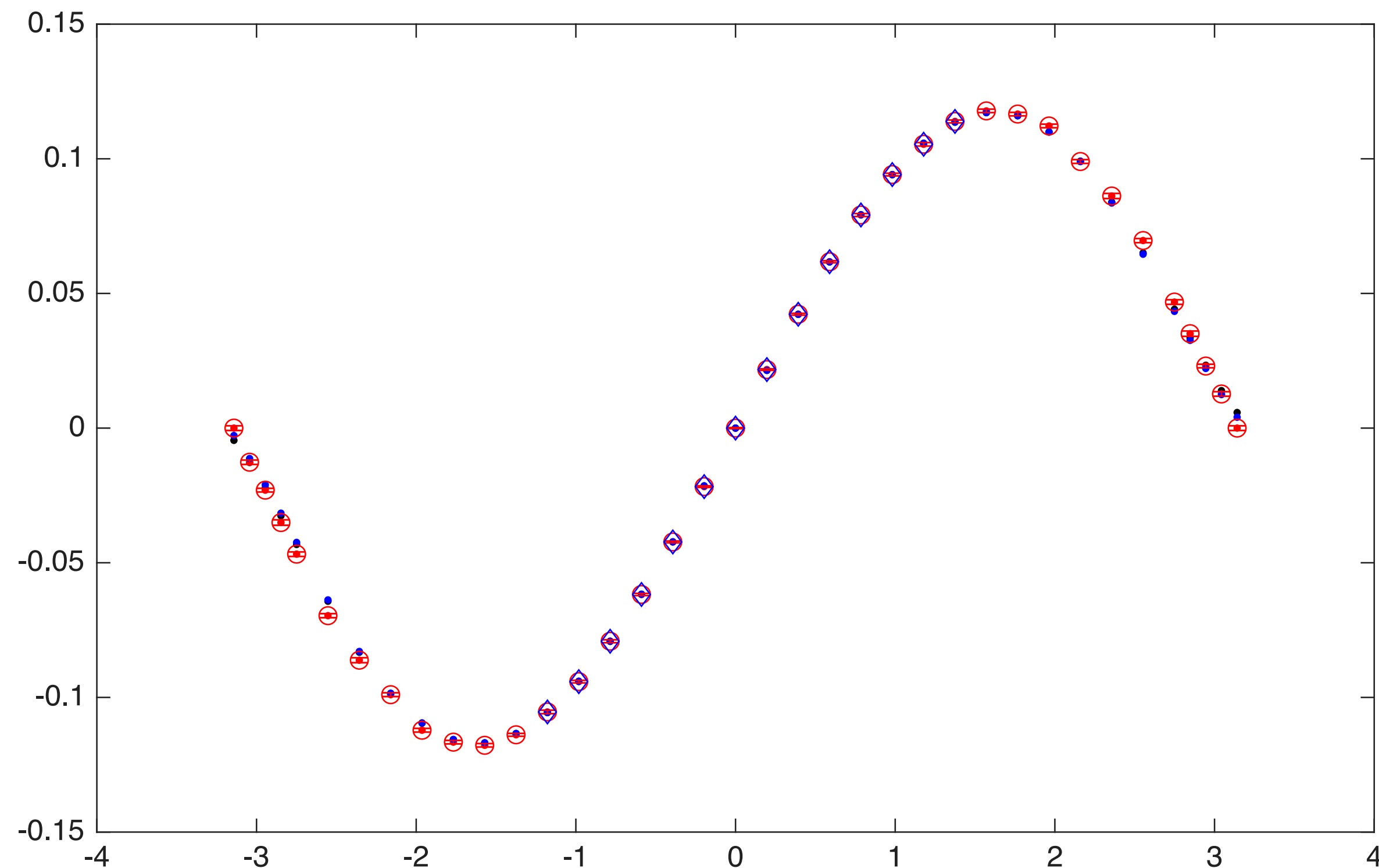


Look! now we are
computing the number density on the real axis
knowing values on the imaginary axis!

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Cross-check (*sanity check...*):

if we now compare **DERIVATIVES in ZERO**, we get
FULL CONSISTENCY with **HotQCD** results

(but with by far less statistics!)

CONCLUSIONS

- This is in the end really taking **analytic properties of QCD** in a fundamental way ...
- ... which means ... as for **LY ZEROS**, we can get extra inputs for Padè ...
- ... but also ... we can look for **when all this machinery** (i.e. this description based on analyticity) **fails**.

