

APPROXFUN AND SOLUTION OF DIFFERENTIAL EQUATIONS IN **JULIA**

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- **ApproxFun** is based on computational Chebyshev and Fourier series
 - Similar to **Chebfun**
- Posing differential equations as **acting on coefficients** results in almost banded operators
 - This allows for solving equations in **infinite dimensions**
 - Doing this efficiently in a general framework depends on **Julia's** multiple dispatch

DEMO I:

BASIC **APPROXFUN** FEATURES

COMPUTATIONAL FOURIER AND CHEBYSHEV SERIES

- The k th Fourier coefficient is

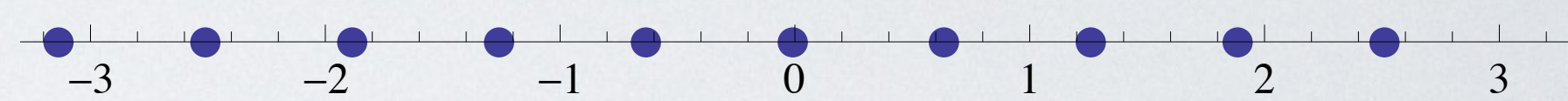
$$\hat{f}_k = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(\theta) e^{-ik\theta} dx$$

- For f periodic and sufficiently smooth, we have the Fourier expansion

$$f(\theta) = \sum_{k=-\infty}^{\infty} \hat{f}_k e^{ik\theta}$$

- Our starting point is to use the FFT to approximate the coefficients \hat{f}_k

- Define the m evenly spaced points on the periodic interval $\boldsymbol{\theta} = (\theta_1, \dots, \theta_m)$:

$$\boldsymbol{\theta} := \left(-\pi, \left(\frac{2}{m} - 1 \right) \pi, \dots, \left(1 - \frac{2}{m} \right) \pi \right) =$$


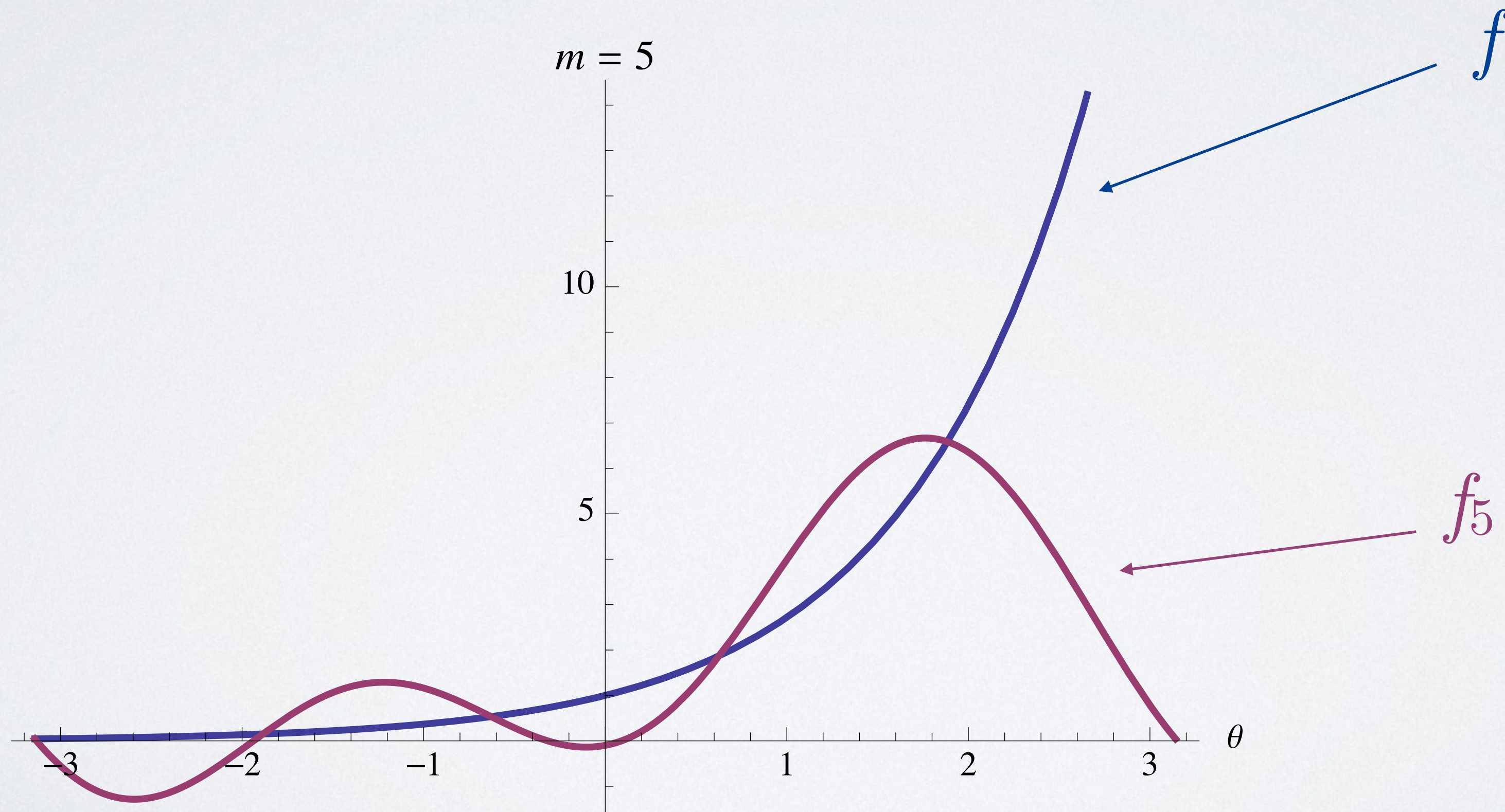
- We can approximate the **Fourier coefficients** using the m point trapezoidal rule

$$\hat{f}_k = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(\theta) d\theta \approx \frac{1}{m} \sum_{j=1}^m f(\theta_j) e^{-k\theta_j} =: \hat{f}_k^m$$

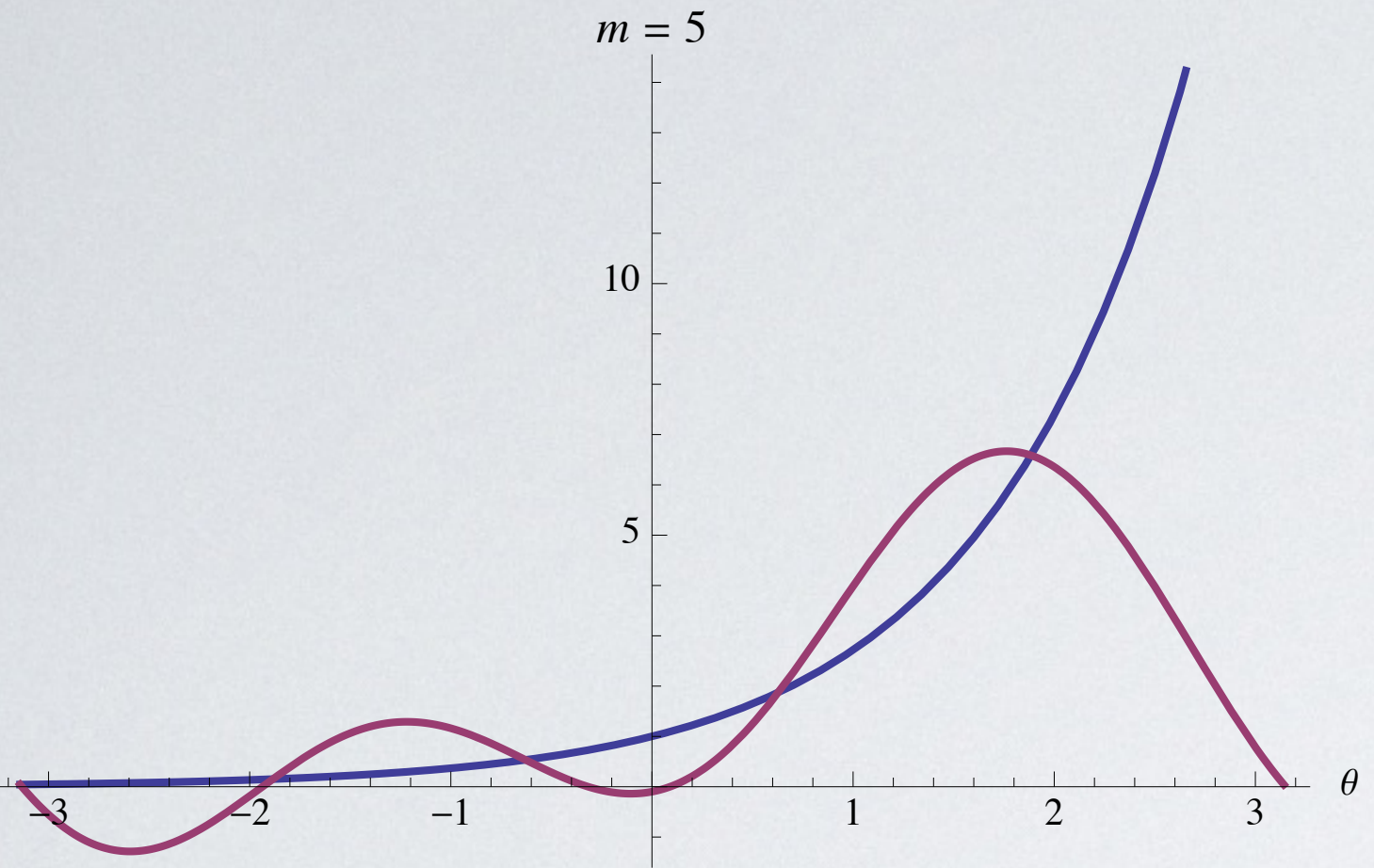
- Using the approximate Fourier coefficients, we get an approximation to f :

$$f_m(\theta) := \begin{cases} \sum_{k=-\frac{m-1}{2}}^{\frac{m-1}{2}} \hat{f}_k^m e^{ik\theta} & m \text{ odd} \\ \frac{\hat{f}_{-m/2}^m}{2} e^{-i\frac{m}{2}\theta} + \sum_{k=1-\frac{m}{2}}^{\frac{m}{2}-1} \hat{f}_k^m e^{ik\theta} + \frac{\hat{f}_{m/2}^m}{2} e^{i\frac{m}{2}\theta} & m \text{ even} \end{cases}$$

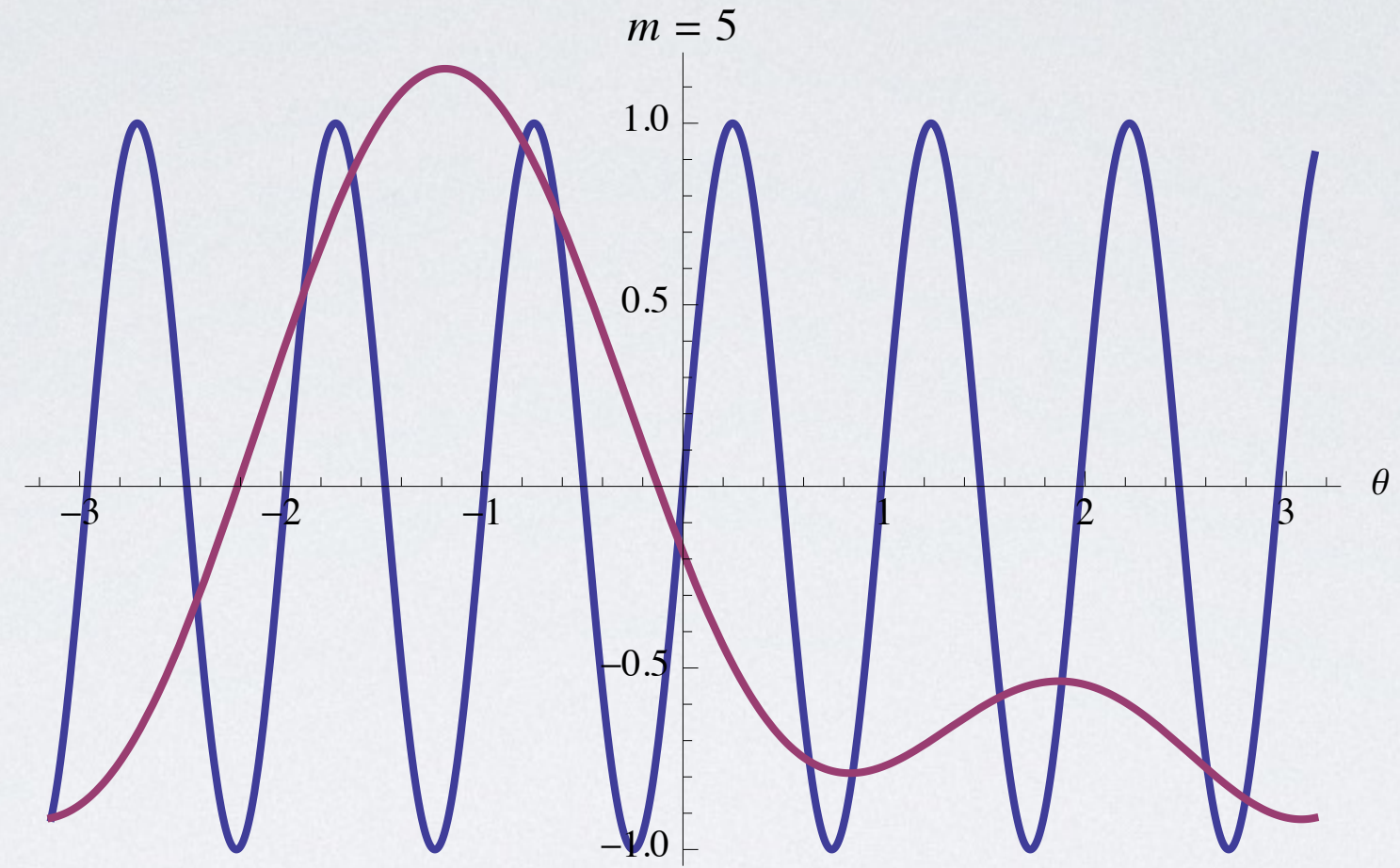
- This interpolates f at θ_j



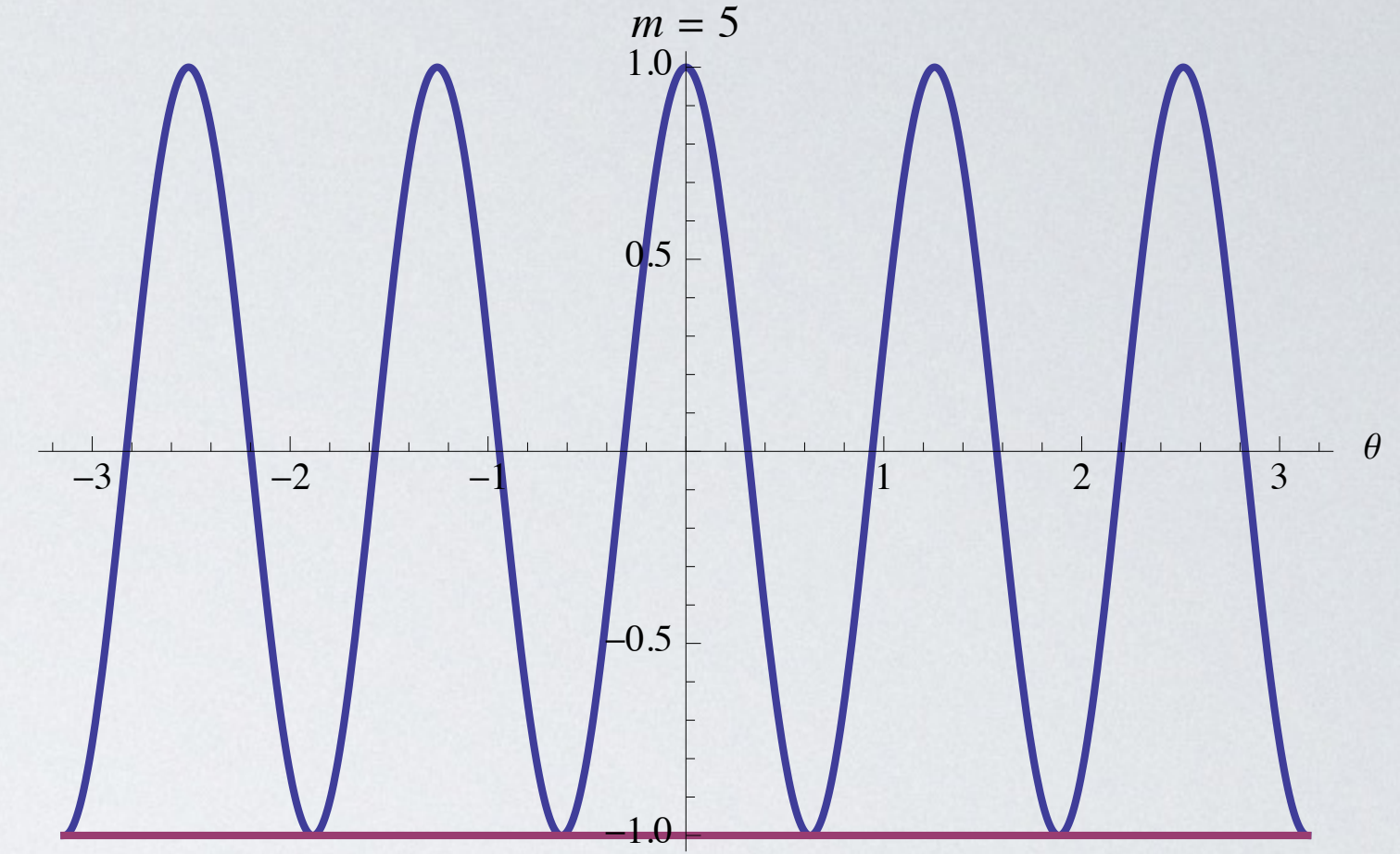
$$e^\theta$$



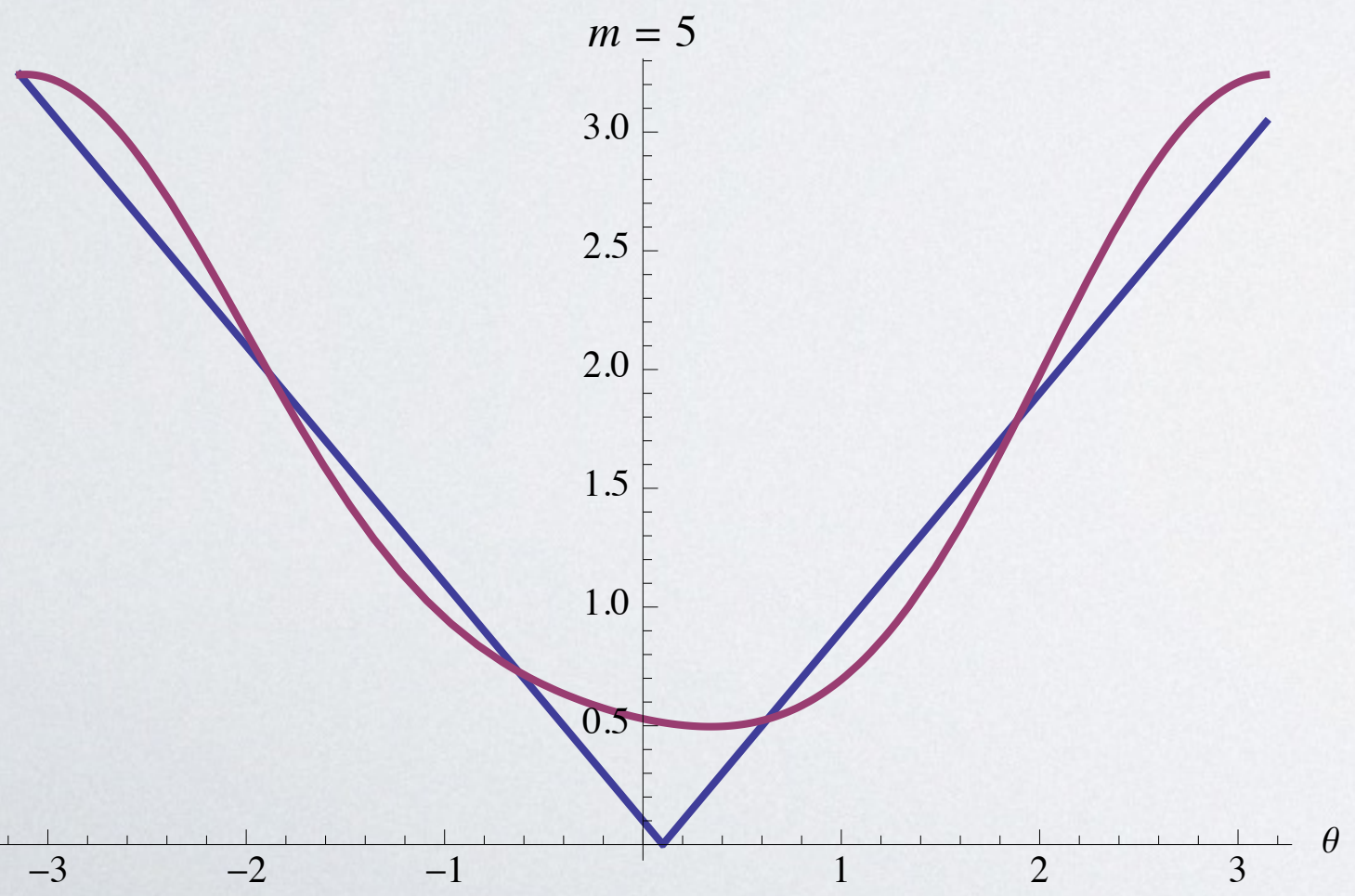
$$\sin \frac{20}{\pi} \theta$$



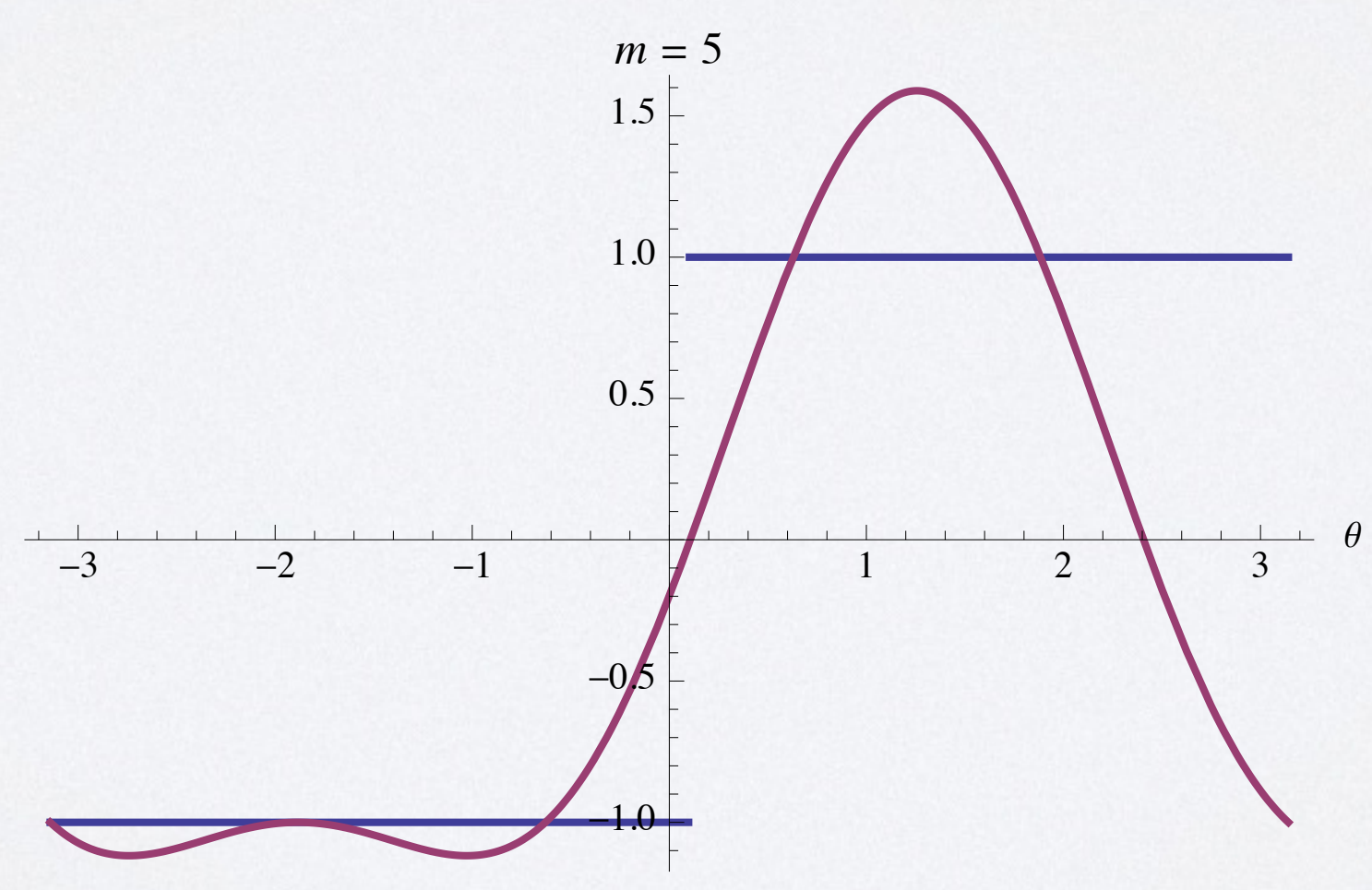
$$\cos 5\theta$$



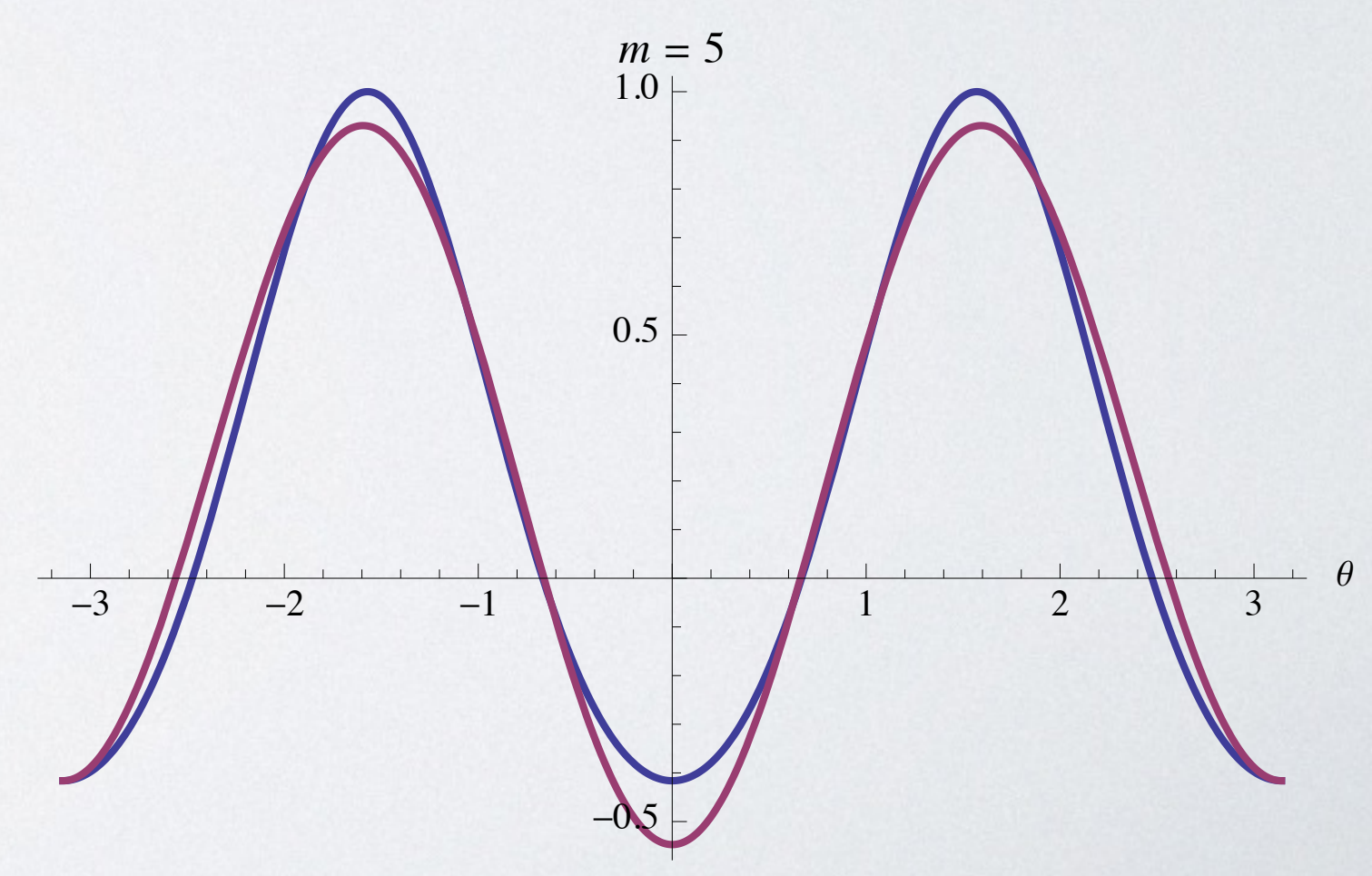
$$|\theta - .1|$$



$$\text{sgn}(\theta - .1)$$

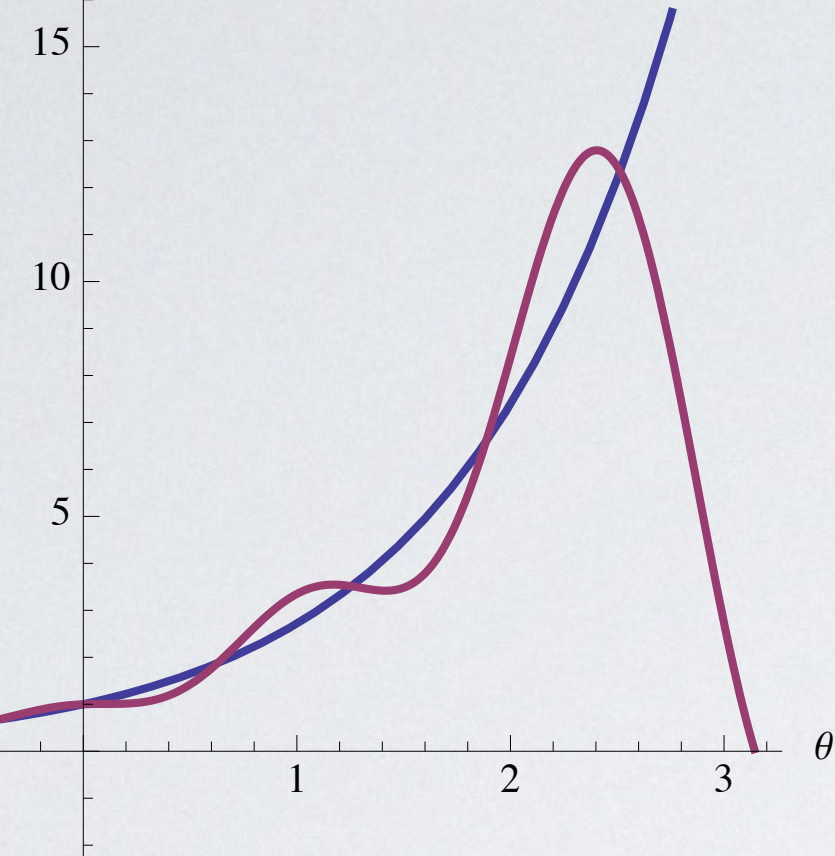


$$\cos 2 \cos \theta$$



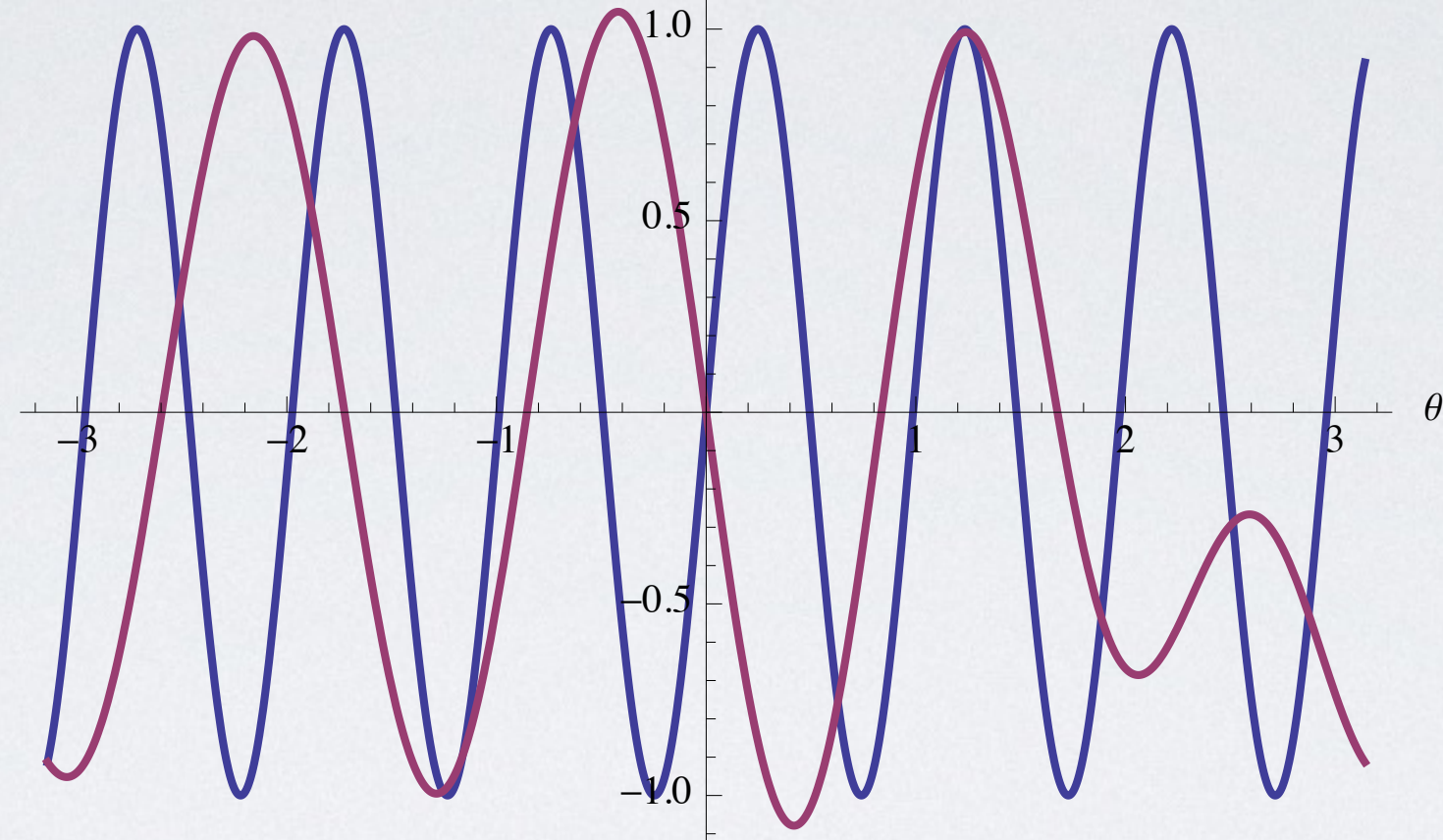
$$e^\theta$$

$m = 10$



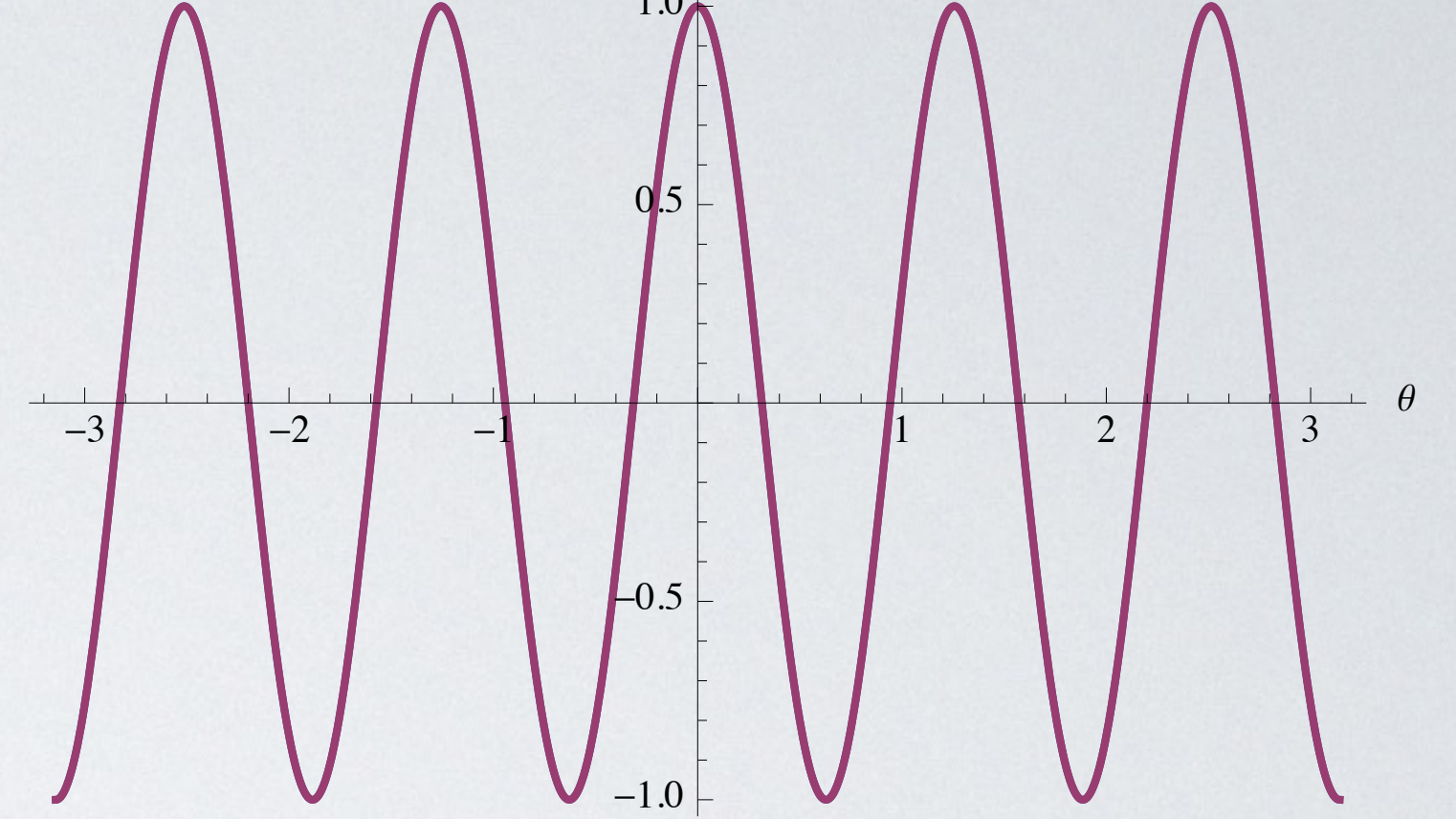
$$\sin \frac{20}{\pi} \theta$$

$m = 10$



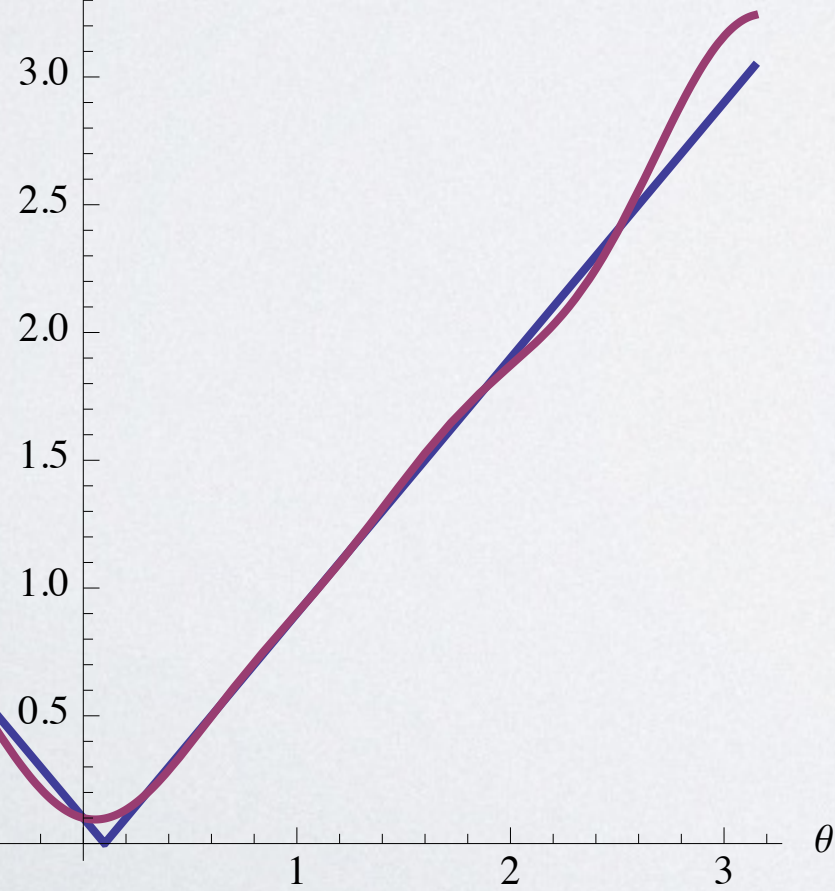
$$\cos 5\theta$$

$m = 10$



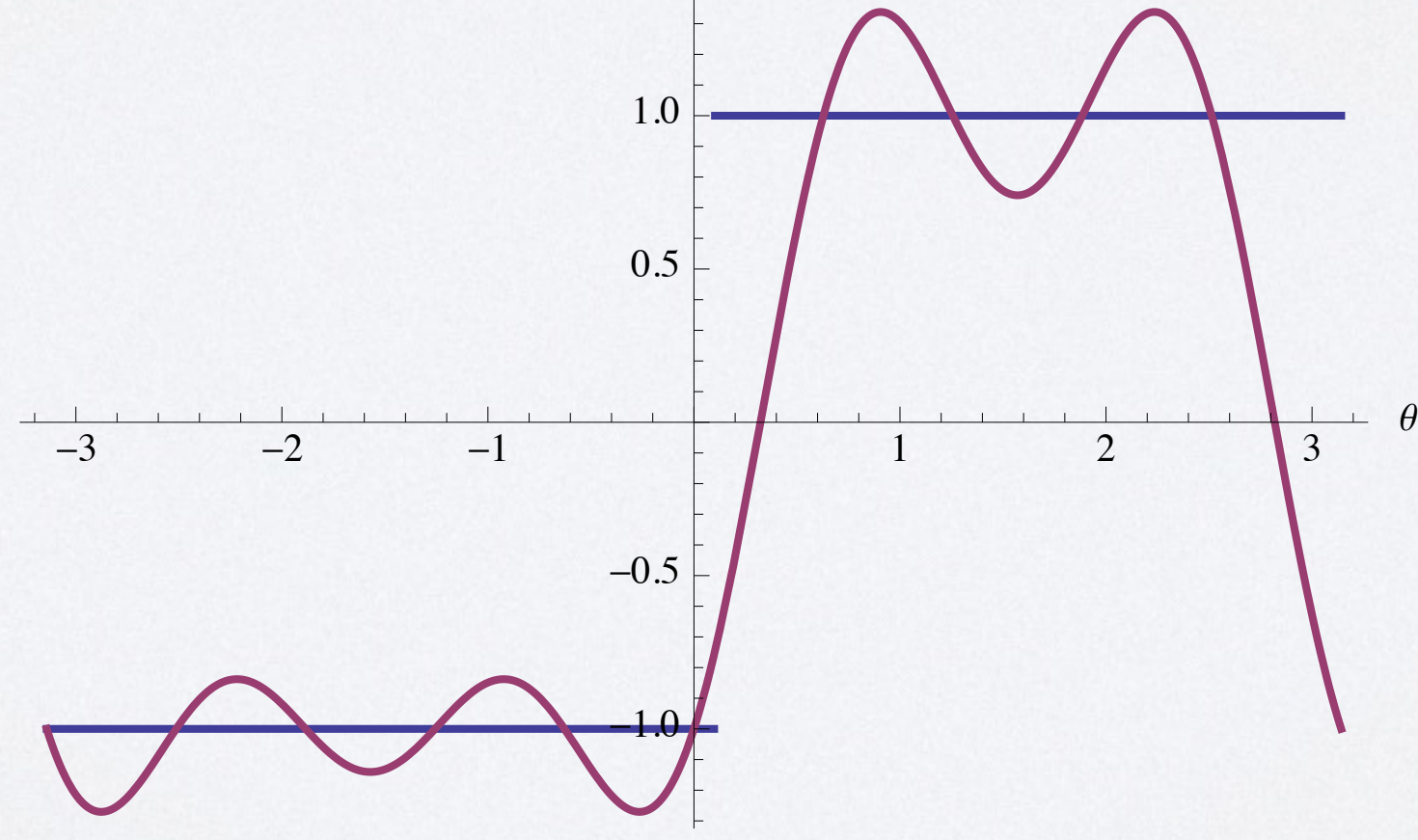
$$|\theta - .1|$$

$m = 10$



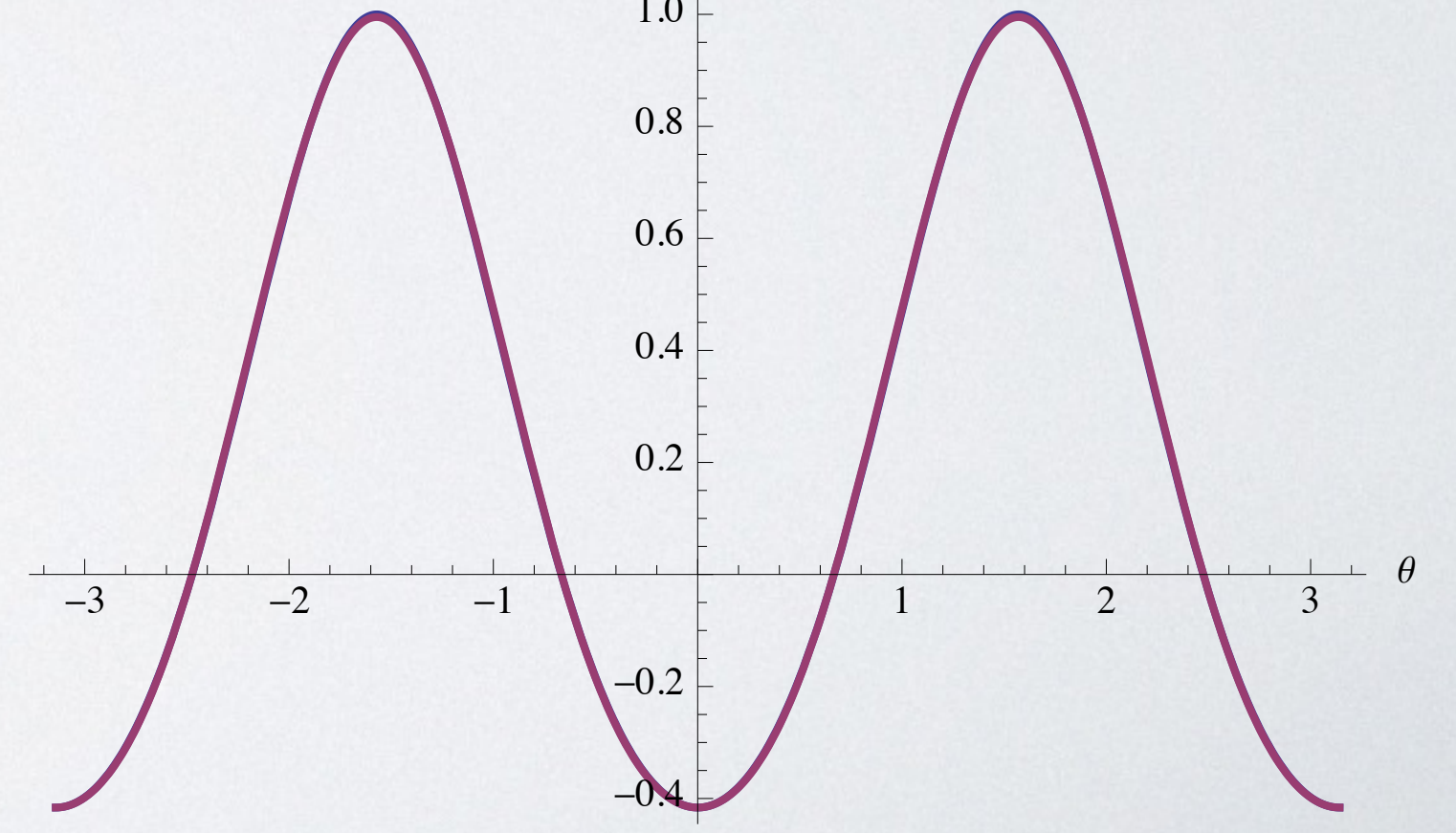
$$\text{sgn}(\theta - .1)$$

$m = 10$



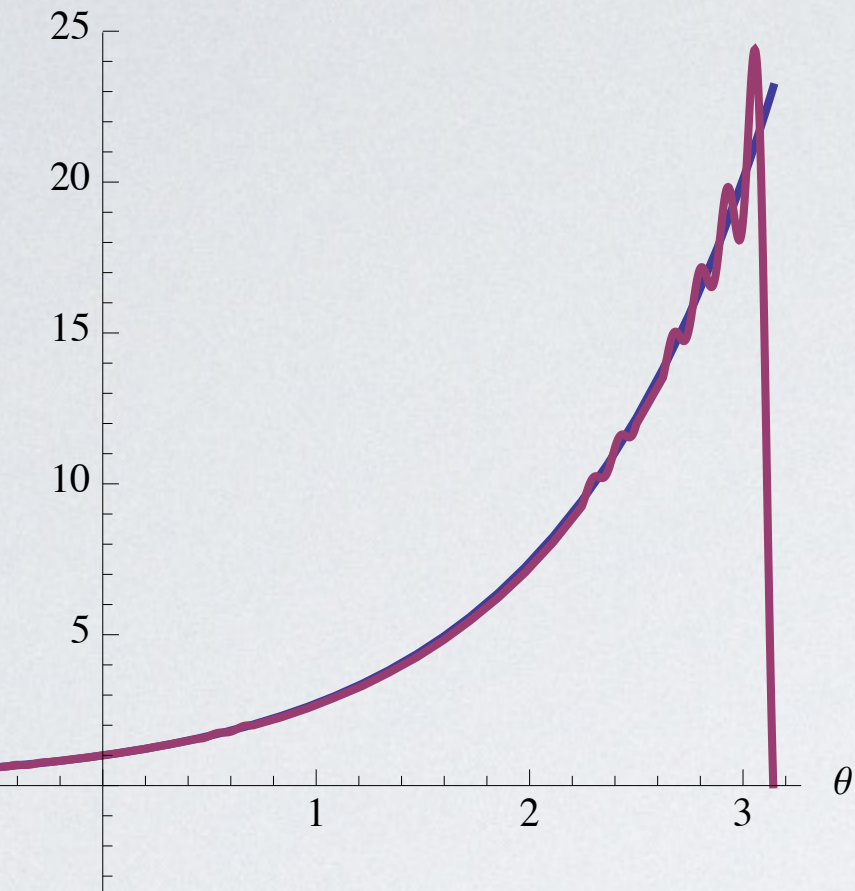
$$\cos 2 \cos \theta$$

$m = 10$



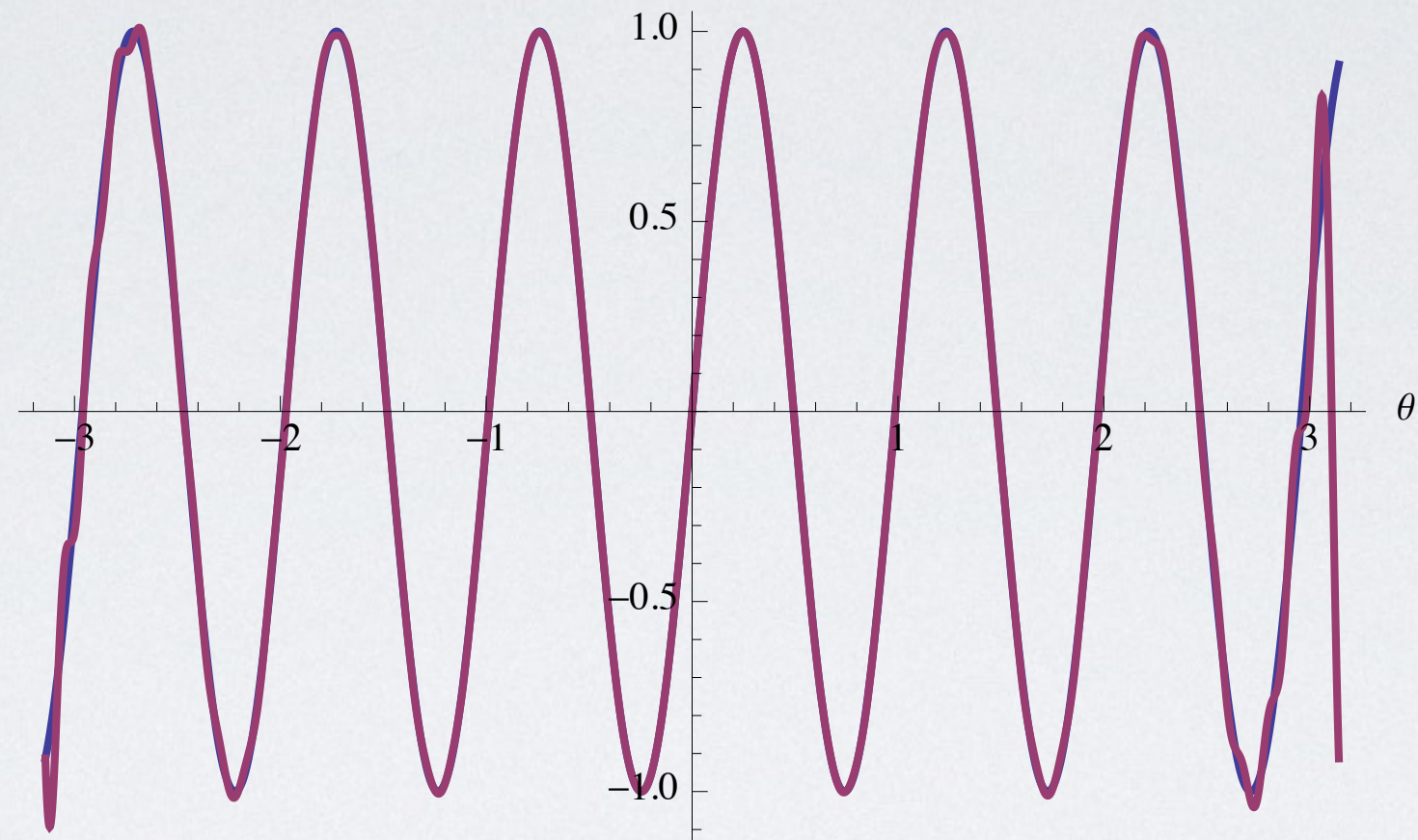
$$e^\theta$$

$m = 100$



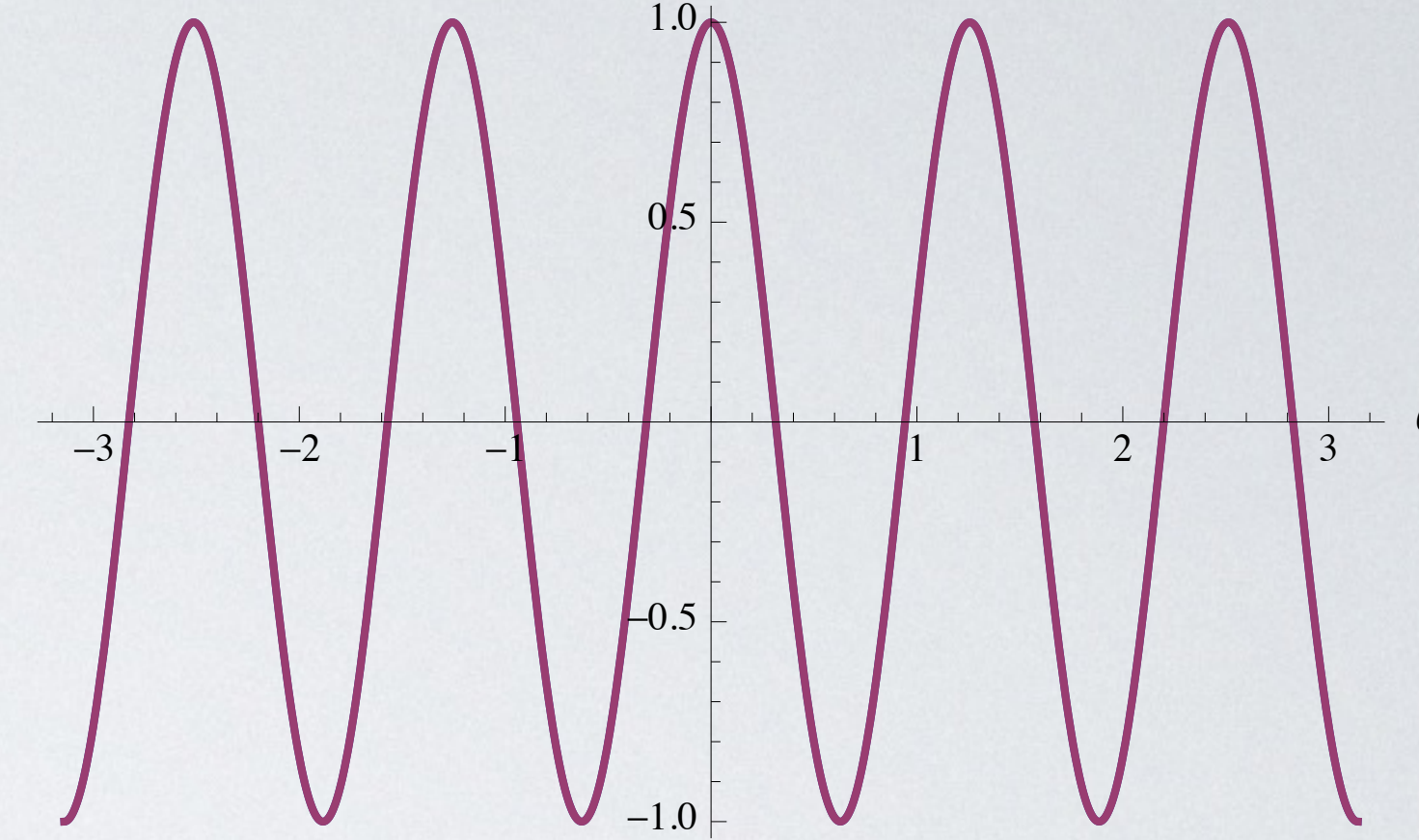
$$\sin \frac{20}{\pi} \theta$$

$m = 100$



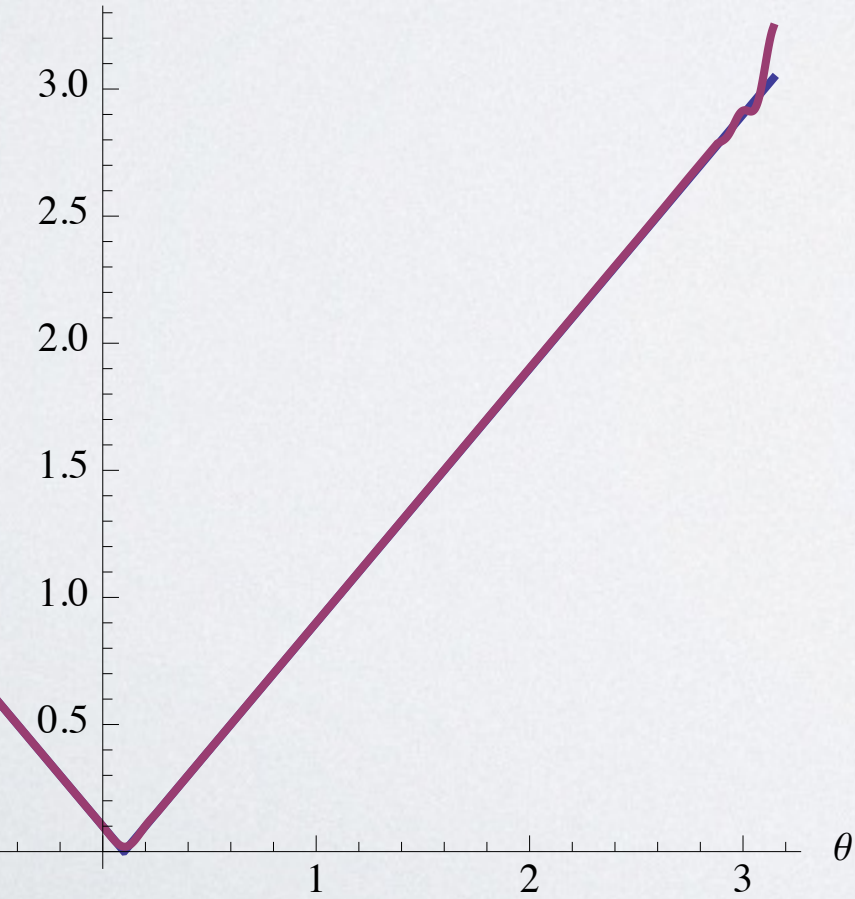
$$\cos 5\theta$$

$m = 100$



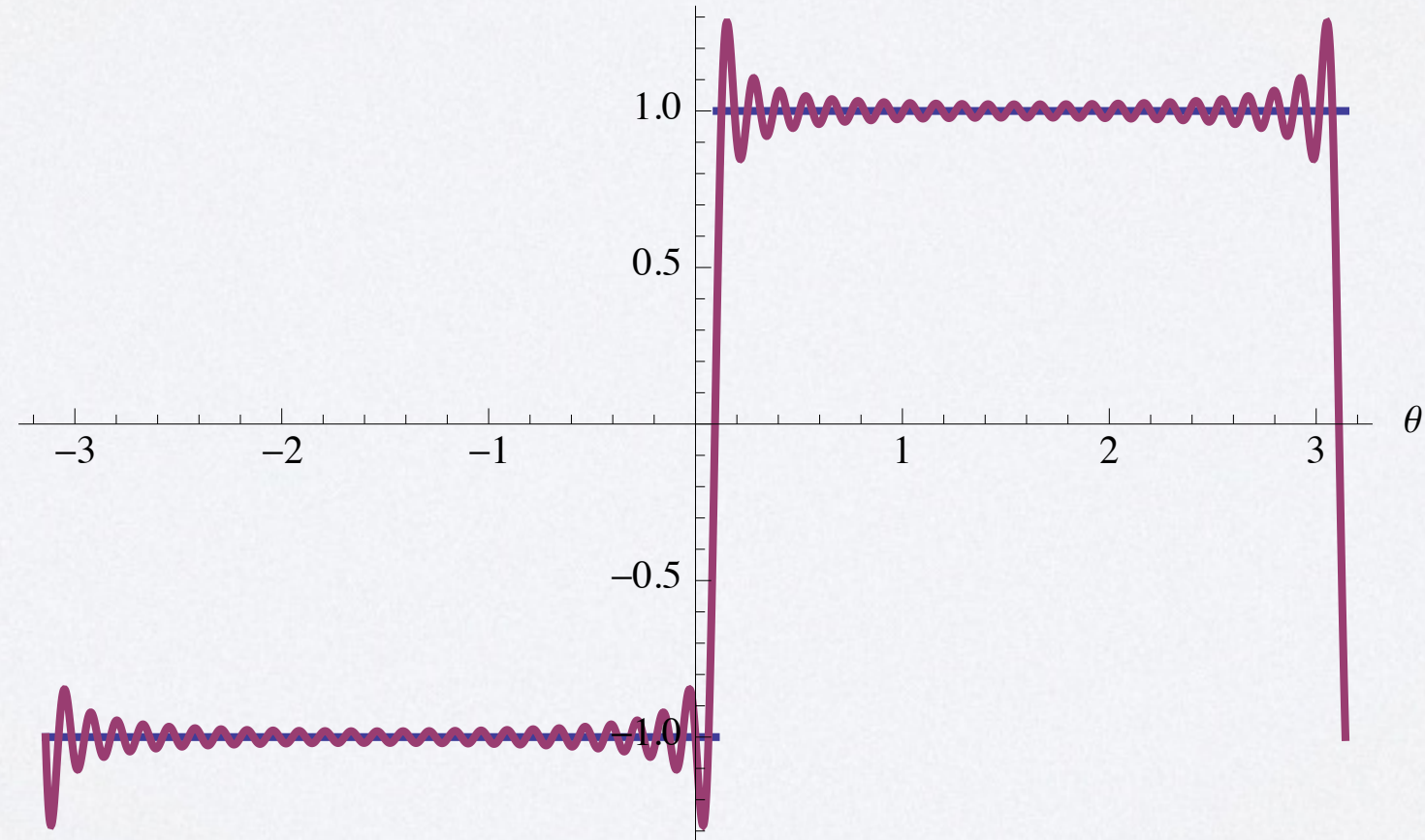
$$|\theta - .1|$$

$m = 100$



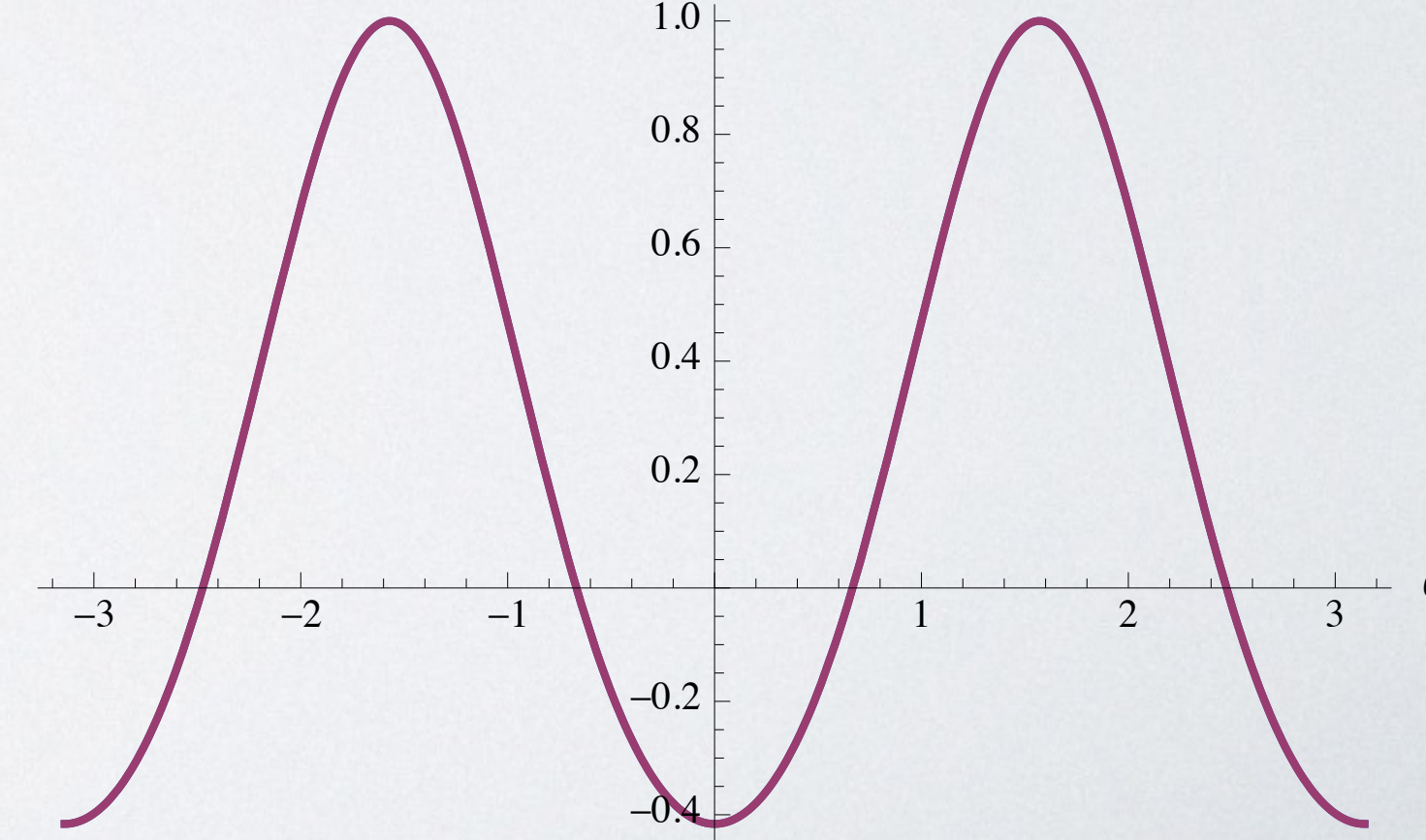
$$\text{sgn}(\theta - .1)$$

$m = 100$



$$\cos 2 \cos \theta$$

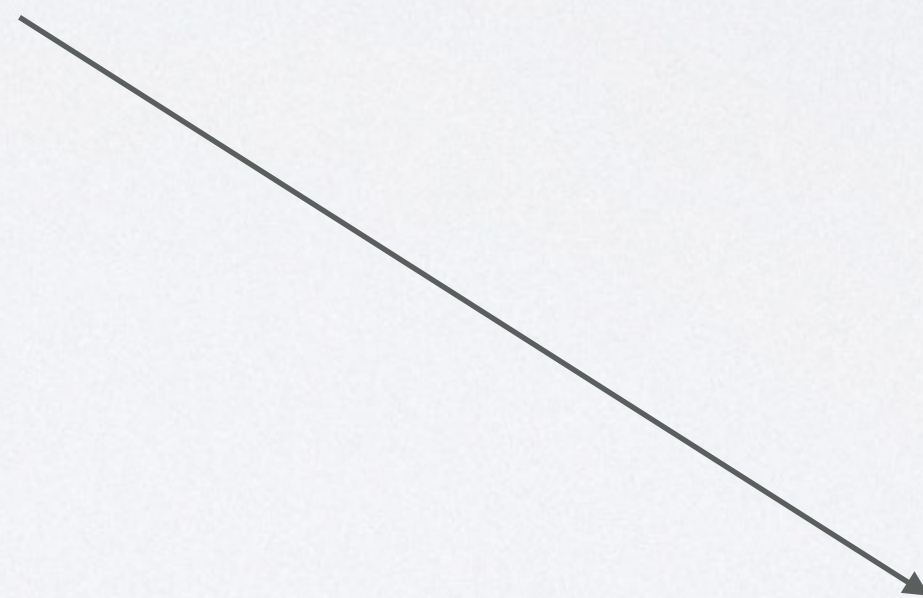
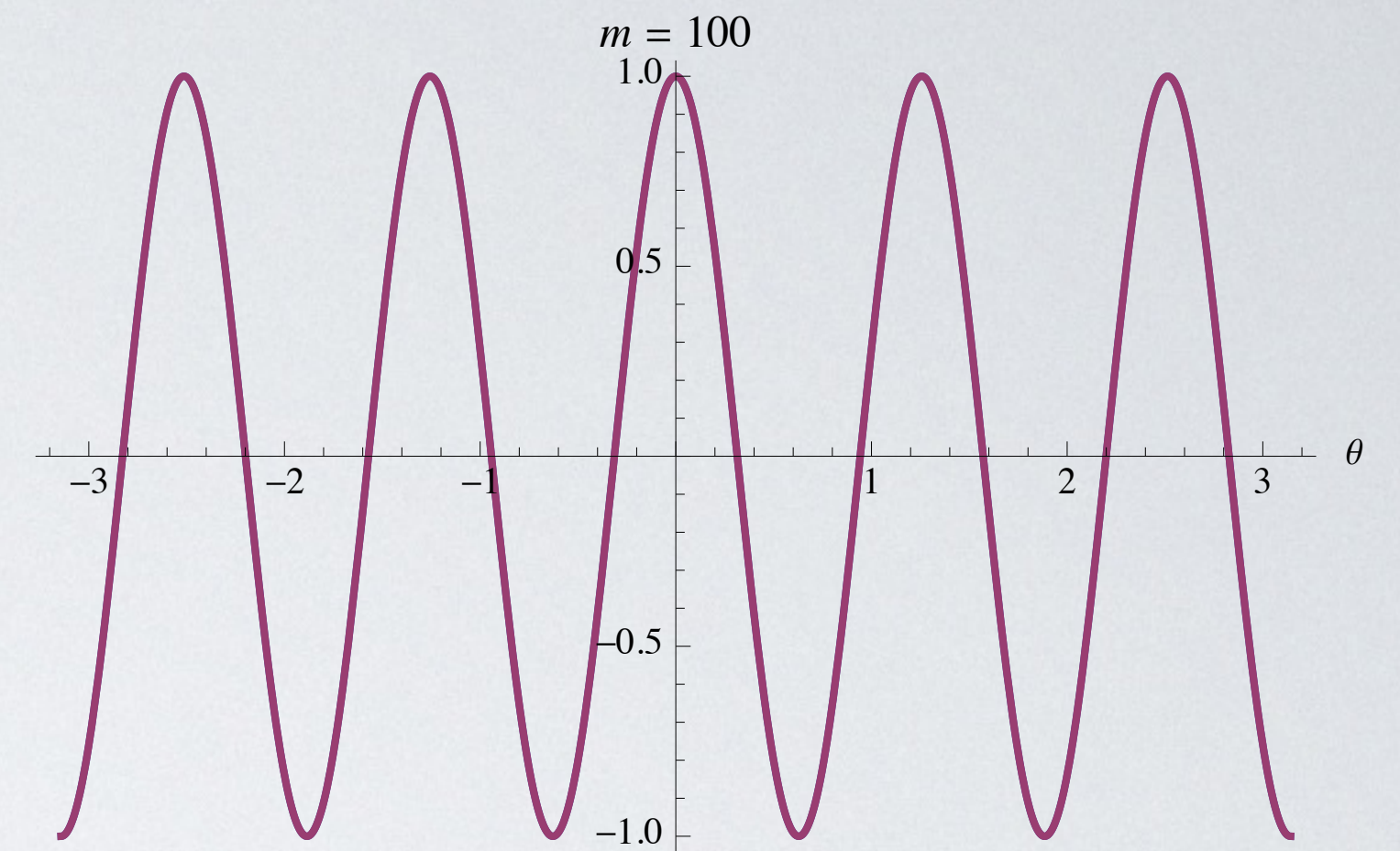
$m = 100$



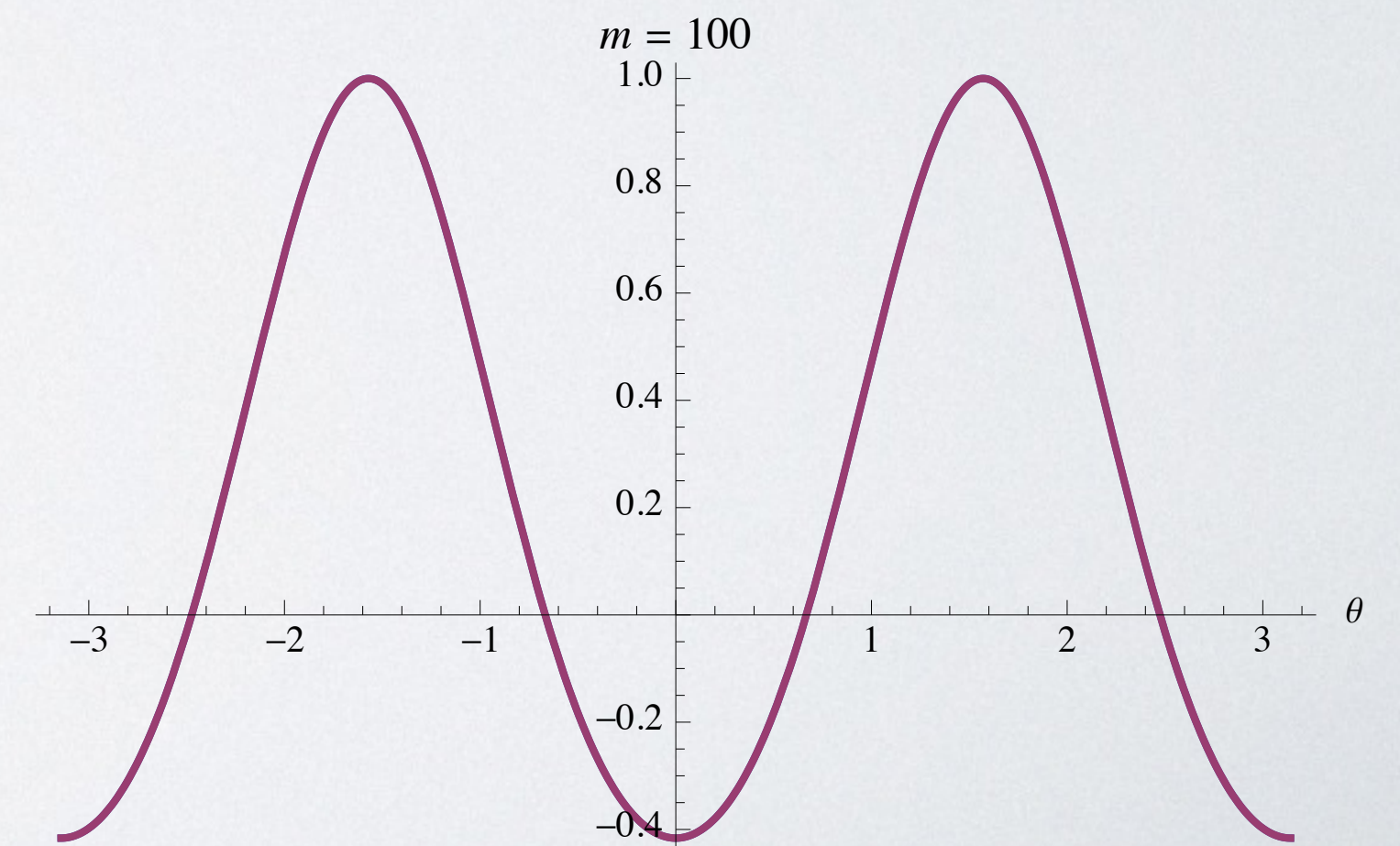
Smooth and periodic converge
very fast!



$$\cos 5\theta$$



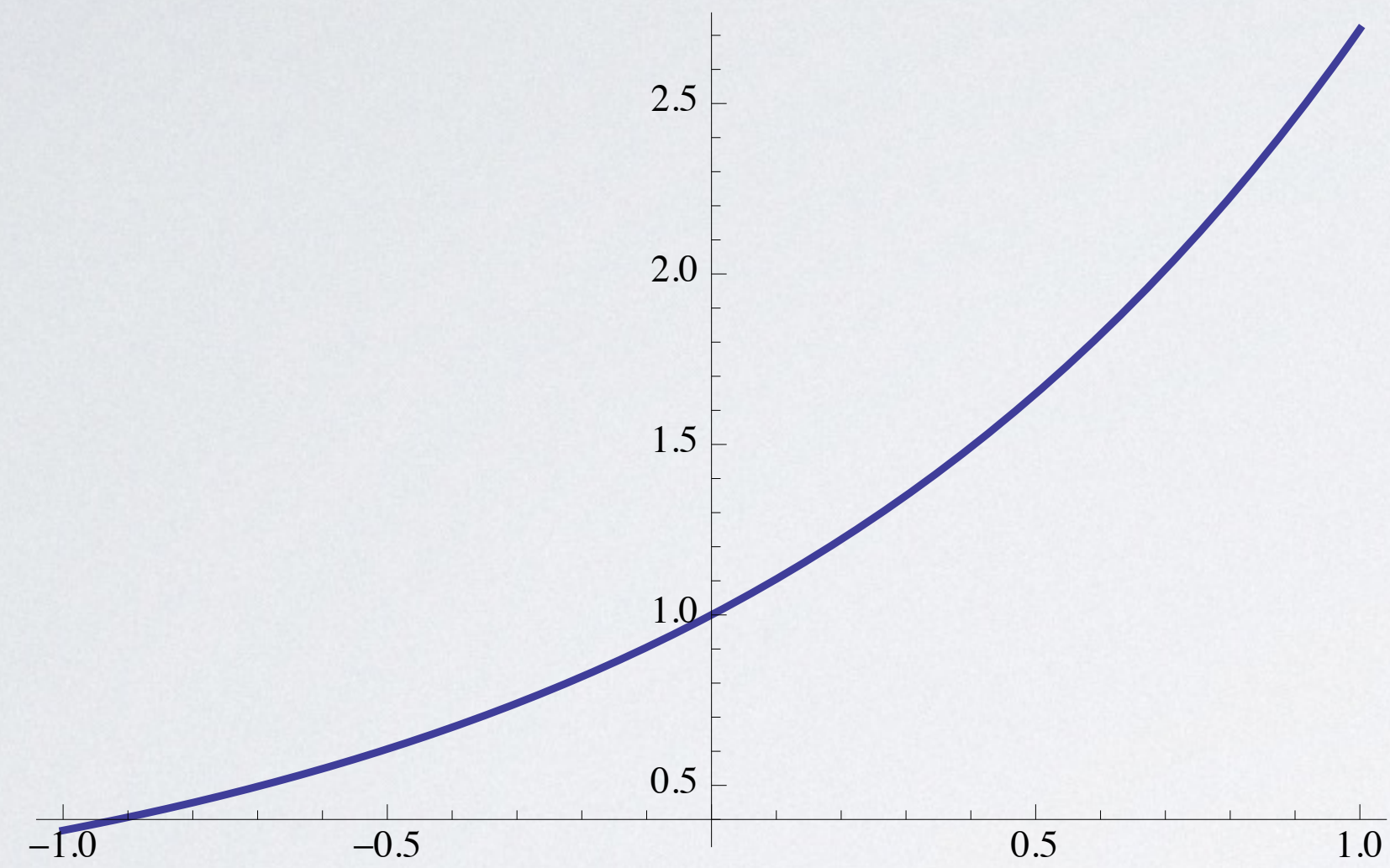
$$\cos 2 \cos \theta$$



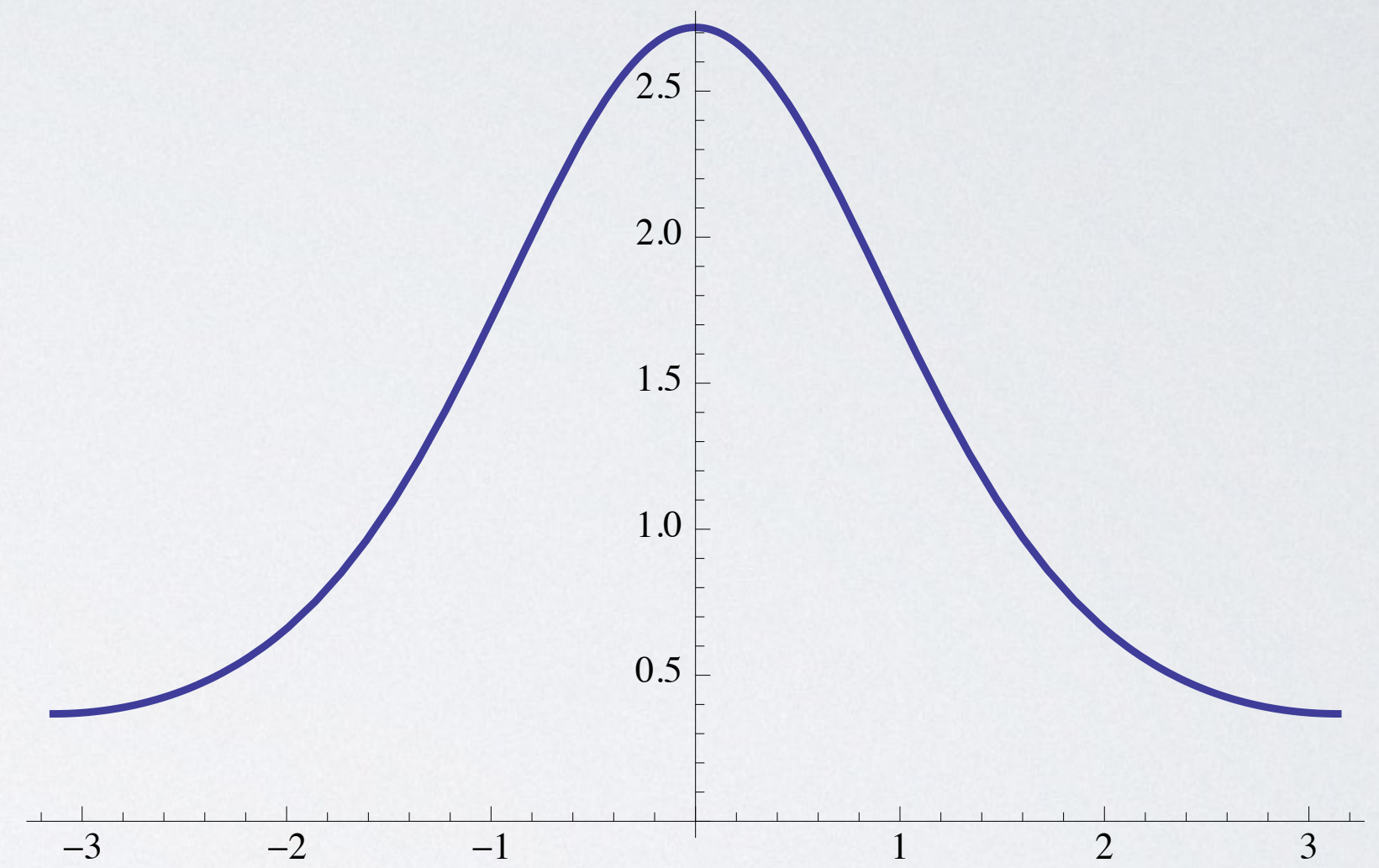
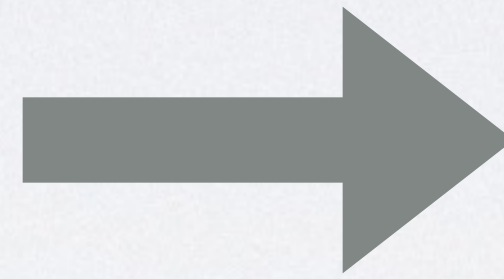
NON-PERIODIC FUNCTIONS

$$e^x$$

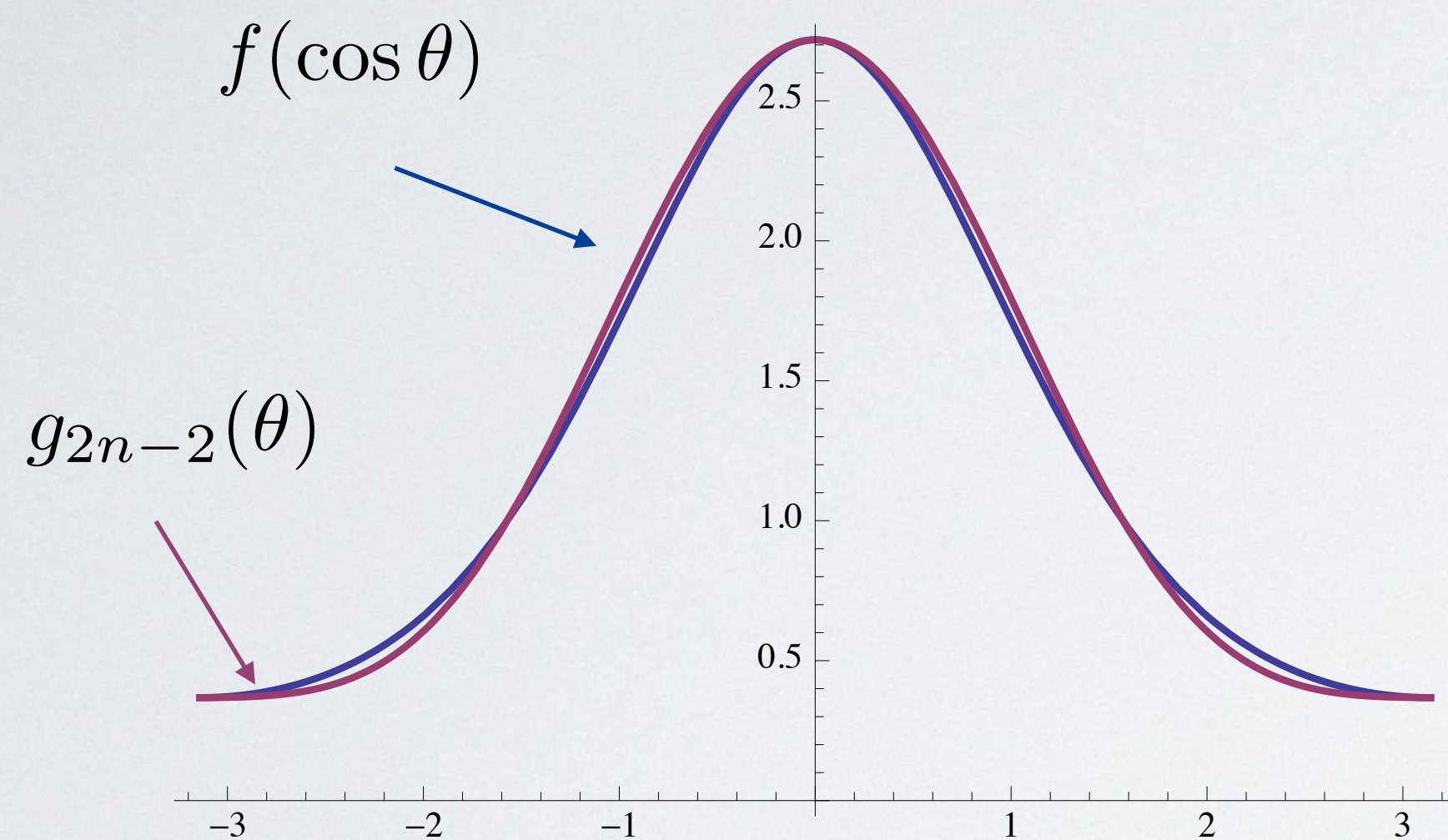
$$e^{\cos \theta}$$



$$x = \cos \theta$$



Smooth and periodic!



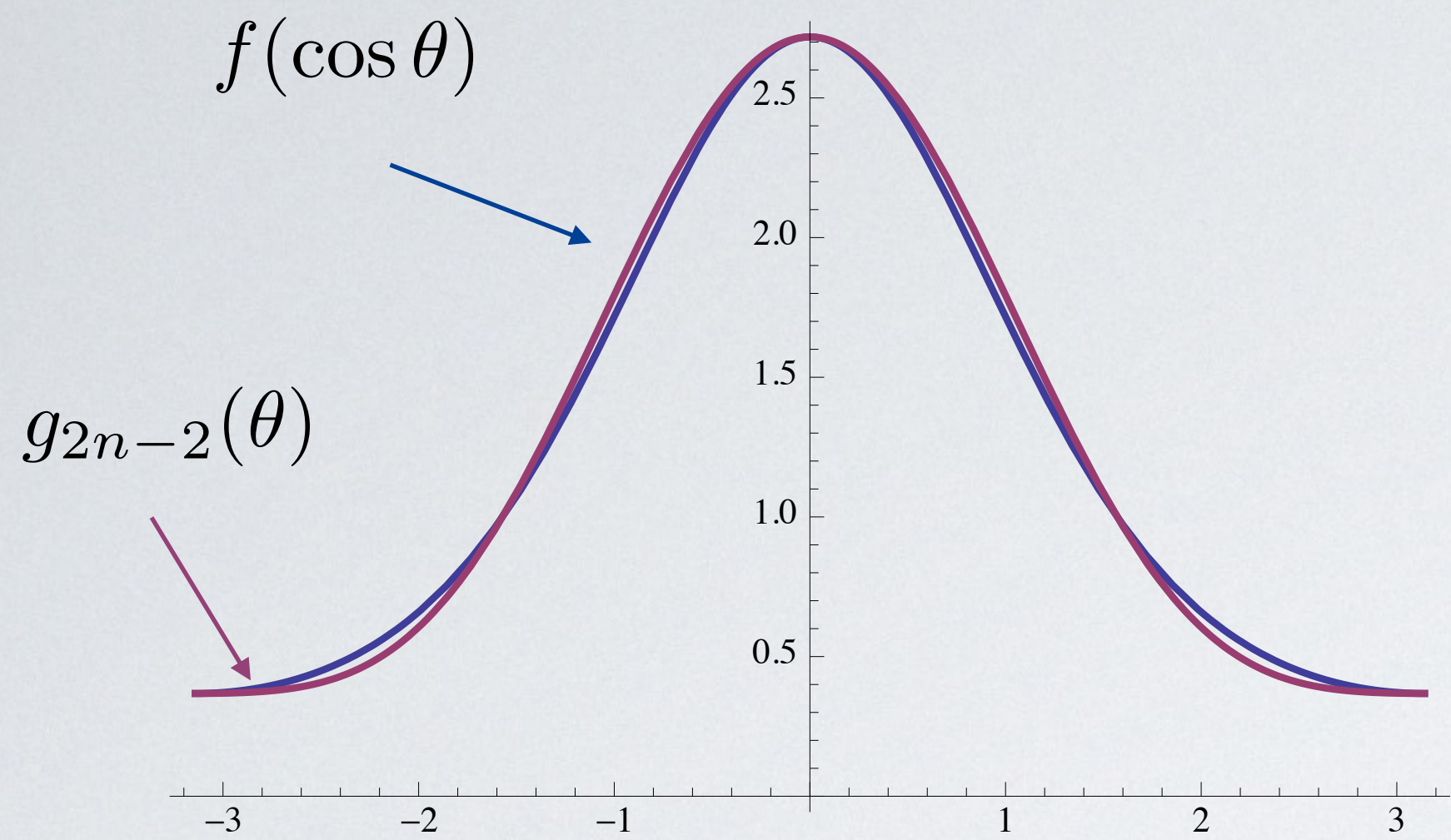
- Let $g(\theta) = f(\cos \theta)$ and approximate

$$g(\theta) \approx g_{2n-2}(\theta)$$

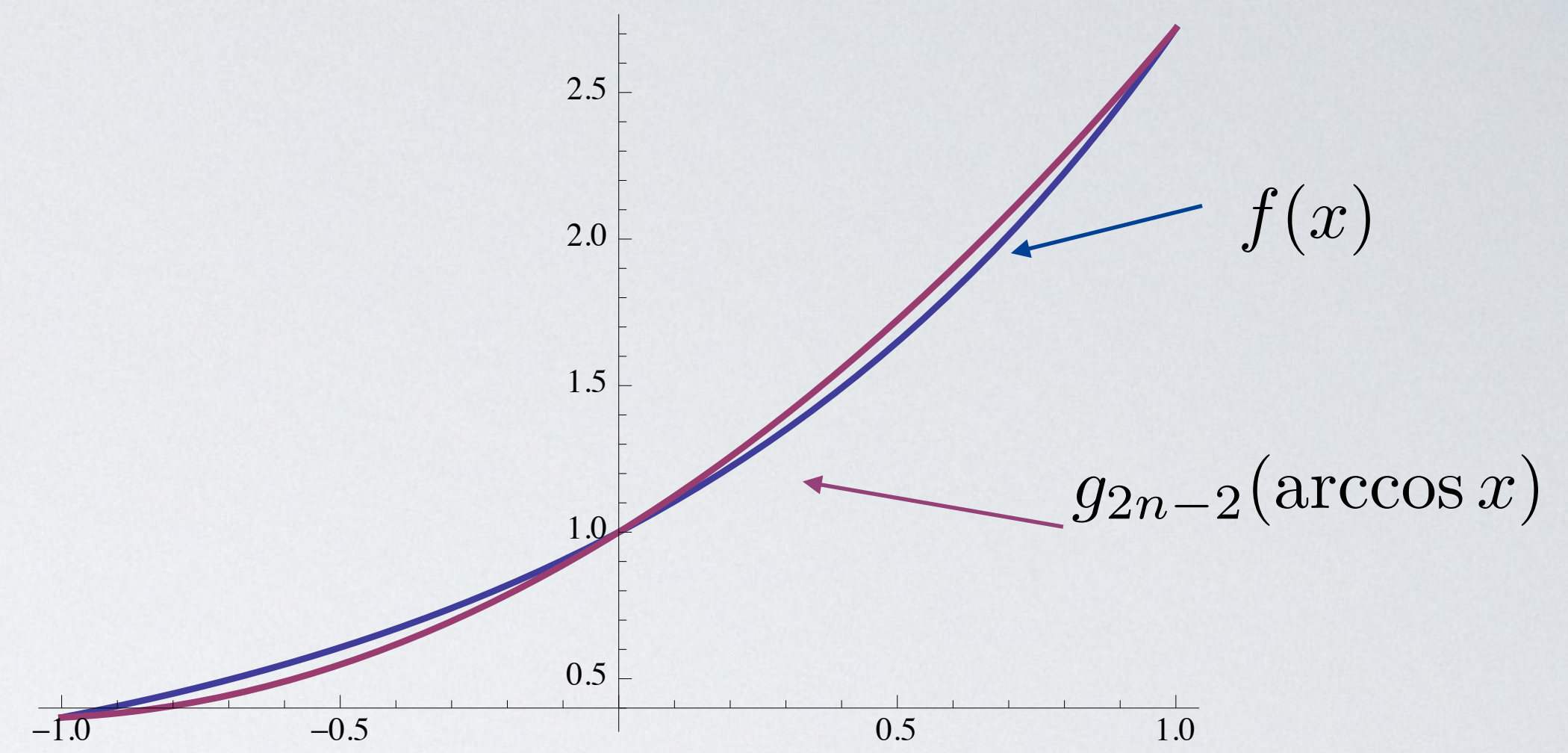
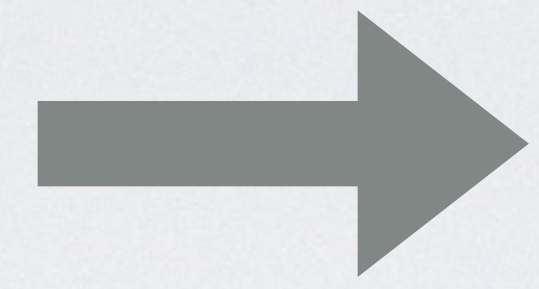
- Symmetry around zero implies that $\hat{g}_k^{2n-2} = \hat{g}_{-k}^{2n-2}$ giving us

$$g_{2n-2}(\theta) = \sum_{k=0}^{n-1} \check{f}_k^n \cos k\theta$$

(where \check{f}_k^n are \hat{g}_k^n but sometimes divided by 2)



$$\theta = \arccos x$$



$$f(x) \approx g_{2n-2}(\arccos x) = \sum_{k=0}^{n-1} \check{f}_k^n T_k(x) \quad \text{for} \quad T_k(x) := \cos k \arccos x$$

- We can replace functions by finite vector of coefficients!

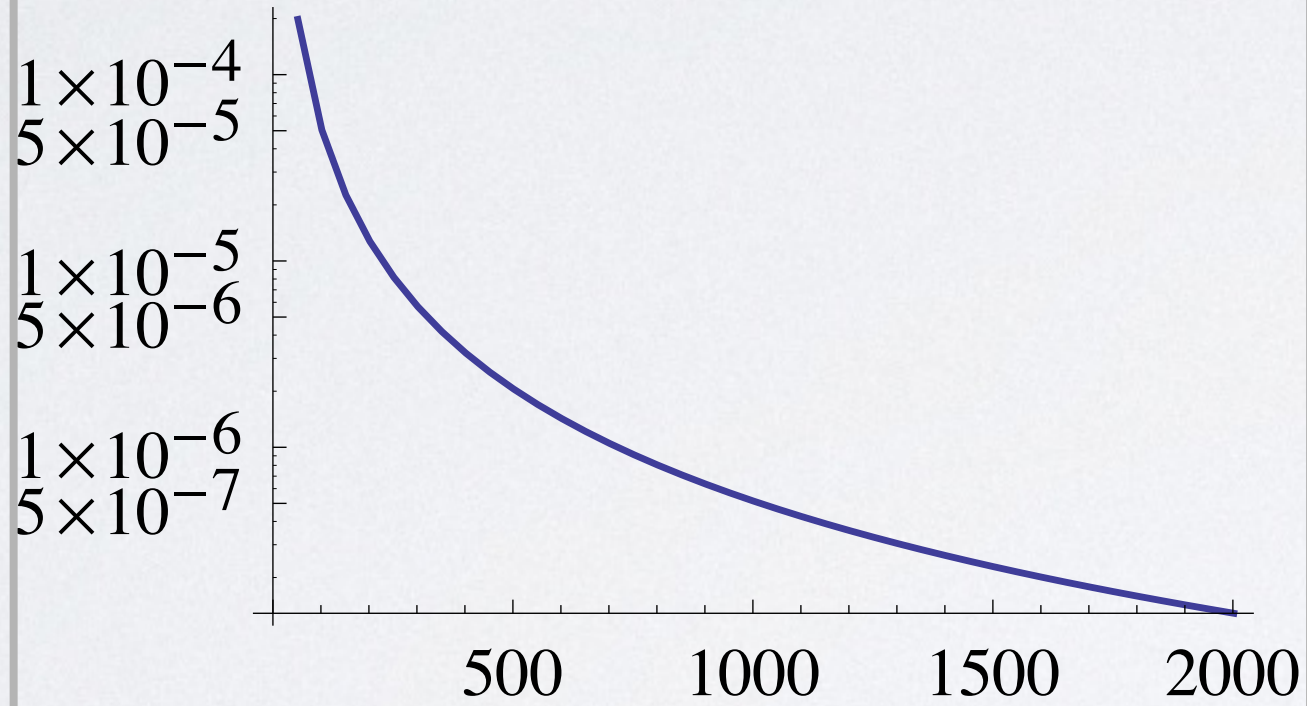
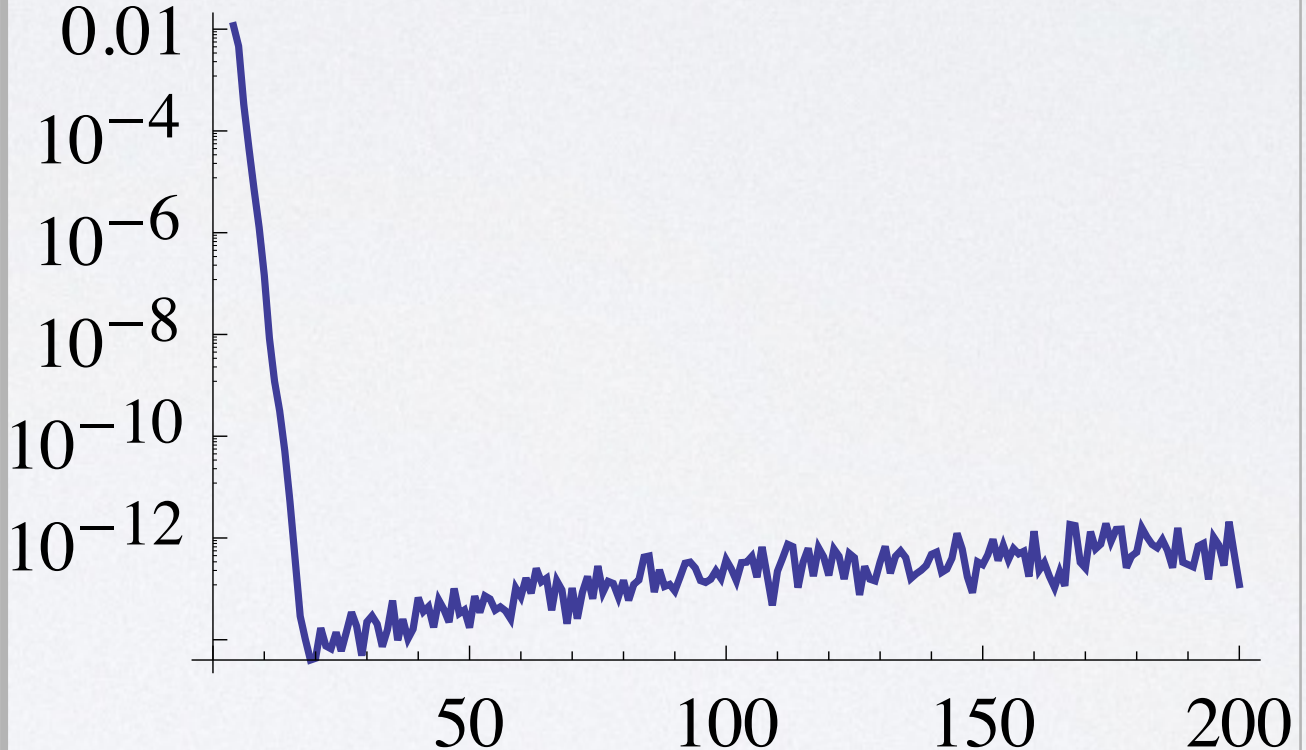
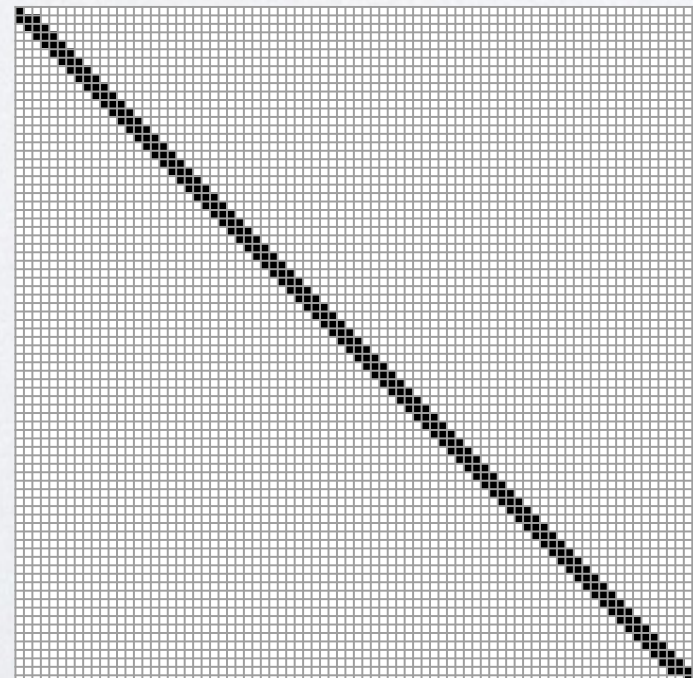
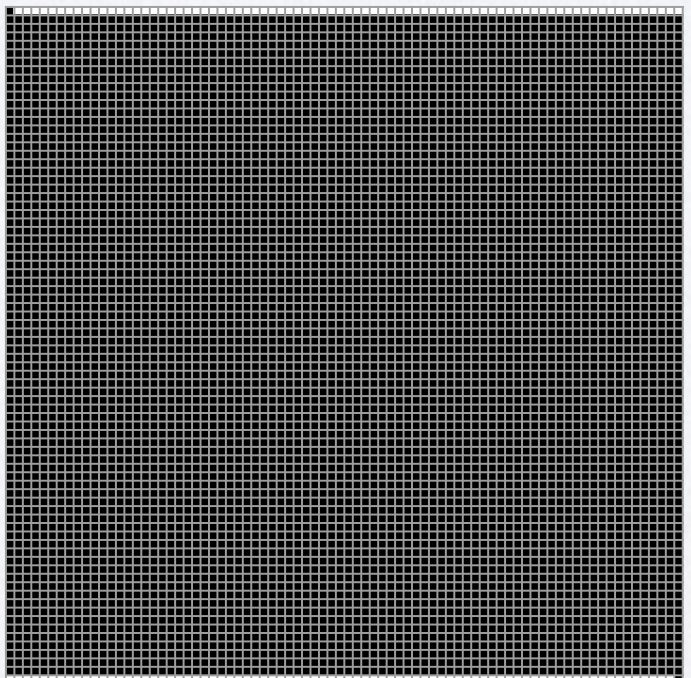
- Thus we represent smooth functions $f(x)$ by a finite-dimensional vector of Chebyshev coefficients

$$f(x) \rightarrow \begin{pmatrix} \check{f}_0^n \\ \vdots \\ \check{f}_{n-1}^n \end{pmatrix}$$

- The length n depends on f , and is chosen automatically by testing tail of coefficients for decay
- Standard function operations then translate to operations on variable length vectors
 - Evaluation is through *Clenshaw's algorithm*
 - Addition is pad to same length and add vectors
 - Multiplication is like a convolution of vectors
- Other operations are accomplished by inverting infinite-dimensional linear equations
 - Solving differential equations, applying special functions and even division

DEMO 2:
DIFFERENTIAL EQUATIONS

- Usual approach to ODEs (finite differences, finite elements, collocation methods): represent differential operator by action at finite number of points
- *Ultraspherical spectral methods* [Olver & Townsend 2013]: represent by action on vector of coefficients, and change basis to get banded operators, and solve in infinite dimensions

	Finite difference	Collocation (<i>Chebfun v4</i>)	Ultraspherical (<i>ApproxFun</i>)
Convergence	 <p>A log-linear plot showing convergence error on the y-axis (ranging from 5×10^{-7} to 1×10^{-4}) versus the number of points on the x-axis (ranging from 0 to 2000). The error decreases as the number of points increases, following a power-law decay.</p>	 <p>A log-linear plot showing convergence error on the y-axis (ranging from 10^{-12} to 0.01) versus the number of points on the x-axis (ranging from 0 to 200). The error drops sharply from 0.01 to below 10^{-12} within the first 25 points and then remains relatively constant with some noise.</p>	
Sparsity	 <p>A matrix sparsity plot showing a clear banded structure, indicating that the differential operator is sparse in the finite difference basis.</p>	 <p>A matrix sparsity plot showing a dense, dark square, indicating that the differential operator is not sparse in the collocation basis.</p>	

- Usual approach to ODEs (finite differences, finite elements, collocation methods): represent differential operator by action at finite number of points
- *Ultraspherical spectral methods* [Olver & Townsend 2013]: represent by action on vector of coefficients, and change basis to get banded operators, and solve in infinite dimensions

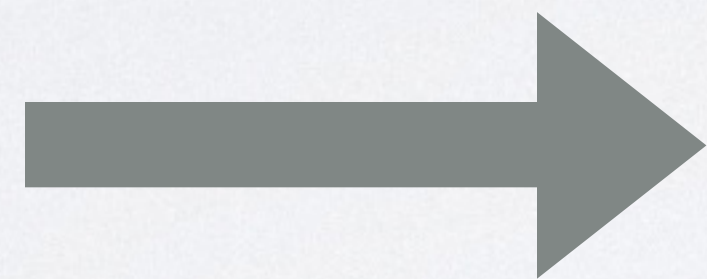
	Finite difference	Collocation (<i>Chebfun v4</i>)	Ultraspherical (<i>ApproxFun</i>)
Convergence			
Sparsity			

1ST ORDER EXAMPLE

$$u(1) = 1$$

$$u' + u = f$$

coefficients




Evaluation at 1

$$\begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 1 & \cdots \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ 1 & 1 & -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} & \cdots \\ & \frac{1}{2} & 2 & 3 & 4 & \cdots & \cdots \\ & & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \frac{1}{2} & \cdots \\ & & & & & & \cdots \\ & & & & & & \cdots \end{pmatrix} \begin{pmatrix} u_0 \\ u_1 \\ u_2 \\ u_3 \\ u_4 \\ \vdots \end{pmatrix} = \begin{pmatrix} 1 \\ \cdots \\ f_0 \\ \vdots \\ f_{n-1} \\ 0 \\ \vdots \end{pmatrix}$$

Instead of truncating and solving, we will **solve in infinite-dimensions**

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

Introduce zero using
Givens rotation
on first two rows



$$\frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} 1. & 1. & 1. & 1. & 1. & 1. & \dots \\ 1. & 1. & -0.5 & 0. & 0. & 0. & \dots \\ 0 & 0.5 & 2. & -0.5 & 0. & 0. & \dots \\ 0 & 0 & 0.5 & 3. & -0.5 & 0. & \dots \\ 0 & 0 & 0 & 0.5 & 4. & -0.5 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} \mathbf{u} = \begin{pmatrix} 1 \\ 2 \\ 1 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

$$\begin{pmatrix} 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & \dots \\ 0 & 0 & -1.06066 & -0.707107 & -0.707107 & -0.707107 & \dots \\ 0 & 0.5 & 2. & -0.5 & 0. & 0. & \dots \\ 0 & 0 & 0.5 & 3. & -0.5 & 0. & \dots \\ 0 & 0 & 0 & 0.5 & 4. & -0.5 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} \mathbf{u} = \begin{pmatrix} 2.12132 \\ 0.707107 \\ 1 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

Now do *Givens rotation*
on next two rows

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

$$\begin{pmatrix}
 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\
 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & \dots \\
 0 & 0 & 1.06066 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\
 0 & 0 & 0.5 & 3. & -0.5 & 0. & 0. & \dots \\
 0 & 0 & 0 & 0.5 & 4. & -0.5 & 0. & \dots \\
 0 & 0 & 0 & 0 & 0.5 & 5. & -0.5 & \dots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{pmatrix} \mathbf{u} = \begin{pmatrix} 2.12132 \\ 1. \\ -0.707107 \\ 0 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

And so on

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

$$\begin{pmatrix} 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\ 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & \dots \\ 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & \dots \\ 0 & 0 & 0 & 2.41209 & -0.753778 & -0.301511 & -0.301511 & -0.301511 & \dots \\ 0 & 0 & 0 & 0.5 & 4. & -0.5 & 0. & 0. & \dots \\ 0 & 0 & 0 & 0 & 0.5 & 5. & -0.5 & 0. & \dots \\ 0 & 0 & 0 & 0 & 0 & 0.5 & 6. & -0.5 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} u = \begin{pmatrix} 2.12132 \\ 1. \\ -0.639602 \\ 0.301511 \\ 0 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

$$\begin{pmatrix} 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\ 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & \dots \\ 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\ 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & \dots \\ 0 & 0 & 0 & 0 & 4.06973 & -0.428393 & 0.061199 & 0.061199 & 0.061199 & \dots \\ 0 & 0 & 0 & 0 & 0.5 & 5. & -0.5 & 0. & 0. & \dots \\ 0 & 0 & 0 & 0 & 0 & 0.5 & 6. & -0.5 & 0. & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 7. & -0.5 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} u = \begin{pmatrix} 2.12132 \\ 1. \\ -0.639602 \\ 0.295235 \\ -0.061199 \\ 0 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

$$\begin{pmatrix} 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\ 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & 0. & \dots \\ 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\ 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\ 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000228355 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\ 0 & 0 & 0 & 0 & 0 & 5.01493 & -0.503731 & -0.00746269 & -0.00746269 & -0.00746269 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0.5 & 6. & -0.5 & 0. & 0. & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 7. & -0.5 & 0. & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} u = \begin{pmatrix} 2.12132 \\ 1. \\ -0.639602 \\ 0.295235 \\ -0.0607423 \\ 0.00746269 \\ 0 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

$$\begin{pmatrix}
 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\
 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & 0. & 0. & \dots \\
 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\
 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\
 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000228355 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\
 0 & 0 & 0 & 0 & 0 & 5.03979 & 0.0940168 & -0.0570311 & -0.00742587 & -0.00742587 & -0.00742587 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 6.02037 & -0.496793 & 0.000740377 & 0.000740377 & 0.000740377 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 7. & -0.5 & 0. & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 9. & \dots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{pmatrix}
 \mathbf{u} =
 \begin{pmatrix}
 2.12132 \\
 1. \\
 -0.639602 \\
 0.295235 \\
 -0.0607423 \\
 0.00742587 \\
 -0.000740377 \\
 0 \\
 0 \\
 0 \\
 \vdots
 \end{pmatrix}$$

$$u' + u = 2 + x \quad \text{and} \quad u(1) = 1$$

$$\begin{pmatrix}
 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\
 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & \dots \\
 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\
 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\
 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000228355 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\
 0 & 0 & 0 & 0 & 0 & 5.03979 & 0.0940168 & -0.0570311 & -0.00742587 & -0.00742587 & -0.00742587 & -0.00742587 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 6.0411 & 0.0842762 & -0.0406453 & 0.000737837 & 0.000737837 & 0.000737837 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 7.0171 & -0.498346 & -0.0000612783 & -0.0000612783 & -0.0000612783 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & 0. & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 9. & -0.5 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 10. & \dots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{pmatrix} u = \begin{pmatrix}
 2.12132 \\
 1. \\
 -0.639602 \\
 0.295235 \\
 -0.0607423 \\
 0.00742587 \\
 -0.000737837 \\
 0.0000612783 \\
 0 \\
 0 \\
 0 \\
 \vdots
 \end{pmatrix}$$

- Suppose we wanted to do back substitution (but don't do it yet!)

$$\begin{pmatrix}
 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\
 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & 0. & 0. & 0. & \dots \\
 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\
 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\
 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000228355 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\
 0 & 0 & 0 & 0 & 0 & 5.03979 & 0.0940168 & -0.0570311 & -0.00742587 & -0.00742587 & -0.00742587 & -0.00742587 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 6.0411 & 0.0842762 & -0.0406453 & 0.000737837 & 0.000737837 & 0.000737837 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 7.0171 & -0.498346 & -0.0000612783 & -0.0000612783 & -0.0000612783 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & 0. & 0. & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 9. & -0.5 & 0. & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 10. & -0.5 & \dots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{pmatrix}
 \mathbf{u} =
 \begin{pmatrix}
 2.12132 \\
 1. \\
 -0.639602 \\
 0.295235 \\
 -0.0607423 \\
 0.00742587 \\
 -0.000737837 \\
 0.0000612783 \\
 0 \\
 0 \\
 0 \\
 \vdots
 \end{pmatrix}$$

- What would the forward error be?
 - Precisely the left over term:

$$\begin{pmatrix}
 0. \\
 0. \\
 0. \\
 0. \\
 0. \\
 0. \\
 0. \\
 0.0000612783 \\
 0 \\
 0 \\
 0 \\
 \vdots
 \end{pmatrix}$$

- Thus we know the forward error already, before we have even done the back substitution!

IMPLEMENTATION IN JULIA

- Abstract data type representing infinite-dimensional operators
 - Each operator knows how to write sub-slices to finite-dimensional matrices
- Algebraic operations of operators
 - We have a "**PlusOperator**" and "**TimesOperator**" that contain a list of operators that they add or multiply
 - Thus we build up a **tree** of operators
- Automatic conversion between different coefficient spaces
 - Won't go in to
- Implementation of Chebyshev specific differentiation, multiplication and integration operators

EXAMPLE OPERATOR:
DERIVATIVE OF TAYLOR SERIES ON
CIRCLE OF RADIUS R

$$\begin{pmatrix} 0 & R^{-1} & & & & \\ & & 2R^{-2} & & & \\ & & & 3R^{-3} & & \\ & & & & 4R^{-4} & \\ & & & & & \ddots \\ & & & & & & \ddots \end{pmatrix}$$

```
type TaylorDerivativeOperator <: BandedOperator{Float64}
    radius
end
```

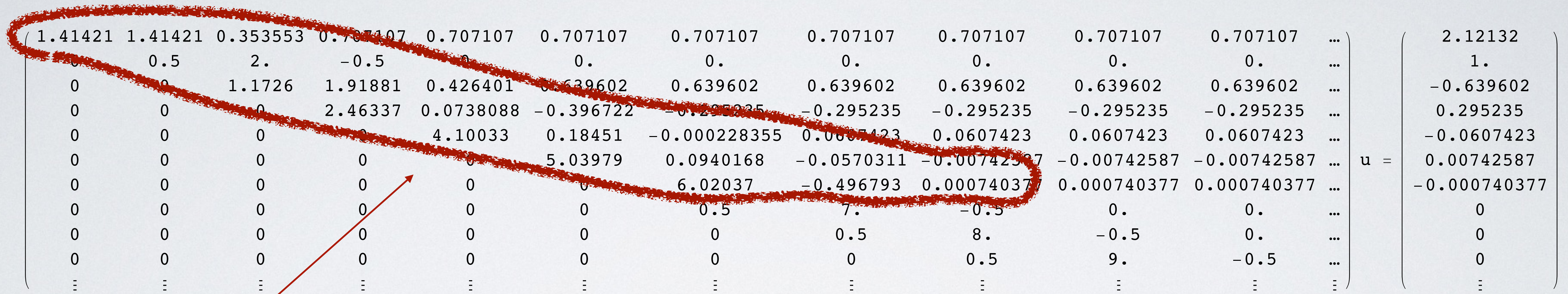
```
ApproxFun.bandrange(::TaylorDerivativeOperator)=0:1 # diagonal and 1 superdiagonal
```

```
function ApproxFun.addentries!(T::TaylorDerivativeOperator, A::ShiftArray, kr::Range1)
    R=T.radius

    for k=max(kr[1],1):kr[end]
        A[k,1]=k/R^k # super-diagonal
    end

    A
end
```


∞ -DIMENSIONAL LINEAR ALGEBRA DATA-TYPE


$$\begin{pmatrix} 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\ 0 & 0.5 & 2. & -0.5 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \dots \\ 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\ 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\ 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000228355 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\ 0 & 0 & 0 & 0 & 0 & 5.03979 & 0.0940168 & -0.0570311 & -0.00742587 & -0.00742587 & -0.00742587 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 6.02037 & -0.496793 & 0.000740377 & 0.000740377 & 0.000740377 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 7. & -0.5 & 0. & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 9. & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} \mathbf{u} = \begin{pmatrix} 2.12132 \\ 1. \\ -0.639602 \\ 0.295235 \\ -0.0607423 \\ 0.00742587 \\ -0.000740377 \\ 0 \\ 0 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

Banded array

∞ -DIMENSIONAL LINEAR ALGEBRA DATA-TYPE

$$\begin{pmatrix}
 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\
 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & 0. & 0. & \dots \\
 0 & 0 & 1.1726 & 1.91881 & 0.425401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\
 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.398722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\
 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000740377 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\
 0 & 0 & 0 & 0 & 0 & 5.03979 & 0.0940168 & -0.00742587 & -0.00742587 & -0.00742587 & -0.00742587 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 6.02037 & -0.496793 & 0.000740377 & 0.000740377 & 0.000740377 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 7. & -0.5 & 0. & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 9. & \dots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{pmatrix} \mathbf{u} = \begin{pmatrix}
 2.12132 \\
 1. \\
 -0.639602 \\
 0.295235 \\
 -0.0607423 \\
 0.00742587 \\
 -0.000740377 \\
 0 \\
 0 \\
 0 \\
 \vdots
 \end{pmatrix}$$

Fill-in by constant multiple of boundary row

∞ -DIMENSIONAL LINEAR ALGEBRA DATA-TYPE

$$\begin{pmatrix} 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\ 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & 0. & 0. & \dots \\ 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\ 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\ 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000228355 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\ 0 & 0 & 0 & 0 & 0 & 5.03979 & 0.0940168 & -0.0570311 & -0.00742587 & -0.00742587 & -0.00742587 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 6.02027 & -0.496793 & 0.000740377 & 0.000740377 & 0.000740377 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 7. & -0.5 & 0. & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & \dots \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 9. & -0.5 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{pmatrix} \mathbf{u} = \begin{pmatrix} 2.12132 \\ 1. \\ -0.639602 \\ 0.295235 \\ -0.0607423 \\ 0.00742587 \\ -0.000740377 \\ 0 \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

Unmodified banded operator

∞ -DIMENSIONAL LINEAR ALGEBRA DATA-TYPE

$$\begin{pmatrix}
 1.41421 & 1.41421 & 0.353553 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & 0.707107 & \dots \\
 0 & 0.5 & 2. & -0.5 & 0. & 0. & 0. & 0. & 0. & 0. & 0. & \dots \\
 0 & 0 & 1.1726 & 1.91881 & 0.426401 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & 0.639602 & \dots \\
 0 & 0 & 0 & 2.46337 & 0.0738088 & -0.396722 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & -0.295235 & \dots \\
 0 & 0 & 0 & 0 & 4.10033 & 0.18451 & -0.000228355 & 0.0607423 & 0.0607423 & 0.0607423 & 0.0607423 & \dots \\
 0 & 0 & 0 & 0 & 0 & 5.03979 & 0.0940168 & -0.0570311 & -0.00742587 & -0.00742587 & -0.00742587 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 6.02037 & -0.496793 & 0.000740377 & 0.000740377 & 0.000740377 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 7. & -0.5 & 0. & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 8. & -0.5 & \dots \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0.5 & 9. & \dots \\
 \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots
 \end{pmatrix}
 \mathbf{u} =
 \begin{pmatrix}
 2.12132 \\
 1. \\
 -0.639602 \\
 0.295235 \\
 -0.0607423 \\
 0.00742587 \\
 -0.000740377 \\
 0 \\
 0 \\
 0 \\
 0 \\
 \vdots
 \end{pmatrix}$$

Banded array

Fill-in by constant multiple of boundary row

Unmodified banded operator

WHY JULIA?

- Originally implemented in `C++`, but not easy to construct operators, or use the implementation
- `MATLAB` is `pass by value`: no fast operators writing to matrices
 - Plus I hate `MATLAB`..., and its OOP is ridiculously slow
- `PYTHON` is too slow for linear algebra, and the nature of the data structure doesn't allow for passing off to `C`

DEMO 3:
PARTIAL DIFFERENTIAL EQUATIONS

- We can solve **general** linear PDEs on a rectangle achieving machine precision accuracy
 - The current approach achieves $\mathcal{O}(n^3)$ operations for n^2 unknowns
 - Is there a **fast, spectrally accurate Poisson solver** on a rectangle to get $\mathcal{O}(n^2)$ operations??
- We do not know yet how to do PDEs in an "infinite-dimensional" way
- We also want to be able to scale to many sub-domains
 - Ideally, we would get $\mathcal{O}(mn^3)$ operations for m sub-domains
 - **Julia** will prove critical for large scale problems

JULIA WISH-LIST

- Fast 3D plotting
- iPad support
- A special matrix-type for banded matrices
- A low-level mode
 - Tired of debugging failed type inference
 - The old NumericExtensions **unsafe_view** was twice as fast as **@inbounds**
- A high-level mode that automatically devectorizes
 - I think **@devec** only works for simple assignment loops
- Typed and compiled anonymous functions