Applied Complex Analysis (2021)

1 Lecture 22: Hermite polynomials

This lecture we overview features of Hermite polynomials, some of which also apply to Jacobi polynomials. This includes

- 1. Rodriguez formula
- 2. Approximation with Hermite polynomials
- 3. Eigenstates of Schrödinger equations with a quadratic well

1.1 Rodriguez formula

Because of the special structure of classical orthogonal weights, we have special Rodriguez formulae of the form

$$p_n(x) = \frac{1}{\kappa_n w(x)} \frac{\mathrm{d}^n}{\mathrm{d}x^n} w(x) F(x)^n$$

where w(x) is the weight and $F(x) = (1 - x^2)$ (Jacobi), x (Laguerre) or 1 (Hermite) and κ_n is a normalization constant.

Proposition (Hermite Rodriguez)

$$H_n(x) = (-1)^n e^{x^2} \frac{d^n}{dx^n} e^{-x^2}$$

Proof We first show that it's a degree n polynomial. This proceeds by induction:

$$H_0(x) = e^{x^2} \frac{d^0}{dx^0} e^{-x^2} = 1$$

$$H_{n+1}(x) = -e^{x^2} \frac{d}{dx} \left[e^{-x^2} H_n(x) \right] = 2x H_n(x) - H'_n(x)$$

Orthogonality follows from integration by parts:

$$\langle H_n, p_m \rangle_{\mathrm{H}} = (-1)^n \int_{-\infty}^{\infty} \frac{\mathrm{d}^n \mathrm{e}^{-x^2}}{\mathrm{d}x^n} p_m \mathrm{d}x = \int_{-\infty}^{\infty} \mathrm{e}^{-x^2} \frac{\mathrm{d}^n p_m}{\mathrm{d}x^n} \mathrm{d}x = 0$$

if m < n

Now we just need to show we have the right constant. But we have

$$\frac{\mathrm{d}^n}{\mathrm{d}x^n}[\mathrm{e}^{-x^2}] = \frac{\mathrm{d}^{n-1}}{\mathrm{d}x^{n-1}}[-2x\mathrm{e}^{-x^2}] = \frac{\mathrm{d}^{n-2}}{\mathrm{d}x^{n-2}}[(4x^2 + O(x))\mathrm{e}^{-x^2}] = \dots = \left[(-1)^n 2^n x^n + O(x^{n-1})\right]\mathrm{e}^{-x^2}$$

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Note this tells us the Hermite recurrence: Here we have the simple expressions

$$H'_n(x) = 2nH_{n-1}(x)$$
 and $\frac{\mathrm{d}}{\mathrm{d}x}[e^{-x^2}H_n(x)] = -e^{-x^2}H_{n+1}(x)$

These follow from the same arguments as before since w'(x) = -2xw(x). But using the Rodriguez formula, we get

$$2nH_{n-1}(x) = H'_n(x) = (-1)^n 2x e^{x^2} \frac{d^n}{dx^n} e^{-x^2} + (-1)^n e^{x^2} \frac{d^{n+1}}{dx^{n+1}} e^{-x^2} = 2xH_n(x) - H_{n+1}(x)$$

which means

$$xH_n(x) = nH_{n-1}(x) + \frac{H_{n+1}(x)}{2}$$

1.2 Approximation with Hermite polynomials

Hermite polynomials are typically used with the weight for approximation of functions: on the real line polynomial approximation is unnatural unless the function approximated is a polynomial as otherwise the behaviour at ∞ is inconsistent (polynomials blow up). Thus we can either use

$$f(x) = e^{-x^2} \sum_{k=0}^{\infty} f_k H_k(x)$$

or

$$f(x) = e^{-x^2/2} \sum_{k=0}^{\infty} f_k H_k(x)$$

** Demonstration **

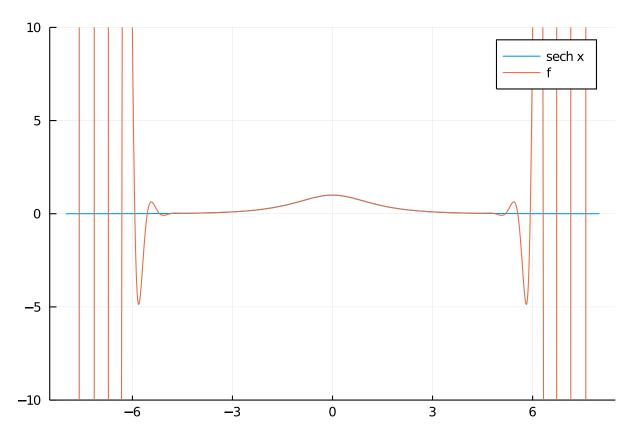
Depending on your problem, getting this wrong can be disasterous. For example, while we can certainly approximate polynomials with Hermite expansions:

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using ApproxFun, Plots
f = Fun(x -> 1+x +x^2, Hermite())
f(0.10)
```

1.10999999999997

We get nonsense when trying to approximate sech (x) by a degree 50 polynomial:

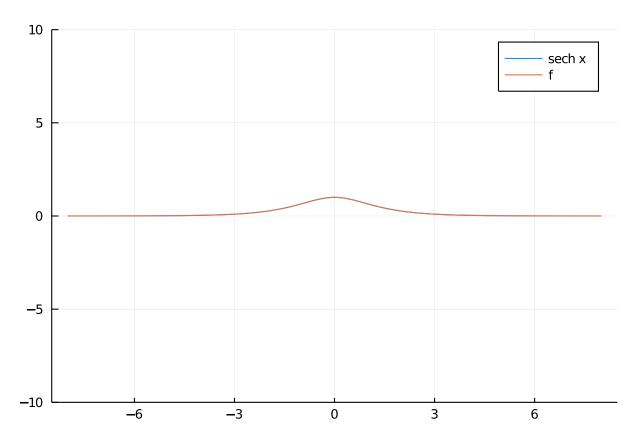
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f = Fun(x -> sech(x), Hermite(), 51)
xx = -8:0.01:8
plot(xx, sech.(xx); ylims=(-10,10), label="sech x")
plot!(xx, f.(xx); label="f")
```



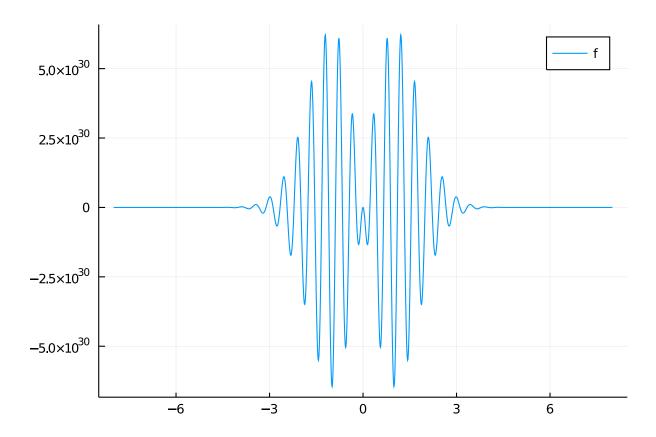
Incorporating the weight $\sqrt{w(x)} = e^{-x^2/2}$ works:

```
f = Fun(x \rightarrow sech(x), GaussWeight(Hermite(),1/2),101)
```

plot(xx, sech.(xx); ylims=(-10,10), label="sech x")
plot!(xx, f.(xx); label="f")



Weighted by $w(x) = e^{-x^2}$ breaks again: $f = Fun(x \rightarrow sech(x), GaussWeight(Hermite()),101)$ plot(xx, sech.(xx); ylims=(-10,10), label="sech x")plot(xx, f.(xx); label="f")



This can be explained by observing that the functions

$$\phi_k(x) = e^{-x^2/2} H_k(x)$$

are orthogonal in $L^2_{\mathbb{R}}$; the functions

$$\widetilde{\phi}_k(x) = e^{-x^2} H_k(x)$$

are orthogonal in $L^2_{\mathbb{R}}(\mathrm{e}^{x^2})$; sech $(x) \in L^2_{\mathbb{R}}$ but sech $(x) \notin L^2_{\mathbb{R}}(\mathrm{e}^{x^2})$. That is, the $\phi_k(x)$ are orthogonal with respect to the weight w(x) = 1 on \mathbb{R} :

$$\langle f, g \rangle_w = \int_{-\infty}^{\infty} f(x)g(x) dx,$$

the $\widetilde{\phi}_k(x)$ are orthogonal with respect to $\widetilde{w} = e^{x^2}$:

$$\langle f, g \rangle_{\widetilde{w}} = \int_{-\infty}^{\infty} f(x)g(x)e^{x^2} dx,$$

for $f(x) = \operatorname{sech}(x)$, we have that

$$\|f\|_w^2 = \langle f, f \rangle_w < \infty, \qquad \|f\|_{\widetilde{w}} = \sqrt{\langle f, f \rangle_{\widetilde{w}}} = \infty.$$

Hence, $\operatorname{sech}(x)$ cannot have a convergent expansion of the form

$$f(x) = \sum_{k=0}^{\infty} f_k \widetilde{\phi}_k(x) = e^{-x^2} \sum_{k=0}^{\infty} f_k H_k(x)$$

in $L^2_{\mathbb{R}}(e^{x^2})$ but it has an expansion in the functions $\phi_k(x)$ in $L^2_{\mathbb{R}}$. Obtaining bounds on the expansion coefficients f_k in the functions $\phi_k(x)$ (as we did previously for Fourier expansions and Chebyshev expansions) is rather complicated and beyond the scope of this module.

1.3 Application: Eigenstates of Schrödinger operators with quadratic potentials

Using the derivative formulae tells us a SturmLiouville operator for Hermite polynomials:

$$e^{x^2} \frac{d}{dx} e^{-x^2} \frac{dH_n}{dx} = 2ne^{x^2} \frac{d}{dx} e^{-x^2} H_{n-1}(x) = -2nH_n(x)$$

or rewritten, this gives us

$$\frac{\mathrm{d}^2 H_n}{\mathrm{d}x^2} - 2x \frac{\mathrm{d}H_n}{\mathrm{d}x} = -2nH_n(x)$$

We therefore have

$$\frac{\mathrm{d}^2}{\mathrm{d}x^2} \left[e^{-\frac{x^2}{2}} H_n(x) \right] = e^{-\frac{x^2}{2}} \left(H_n''(x) - 2x H_n'(x) + (x^2 - 1) H_n(x) \right) = e^{-\frac{x^2}{2}} (x^2 - 1 - 2n) H_n(x)$$

In other words, for the Hermite function $\psi_n(x)$ we have

$$\frac{\mathrm{d}^2\psi_n}{\mathrm{d}x^2} - x^2\psi_n = -(2n+1)\psi_n$$

and therefore ψ_n are the eigenfunctions.

We want to normalize. In Schrödinger equations the square of the wave $\psi(x)^2$ represents a probability distribution, which should integrate to 1. Here's a trick: we know that

$$x \underbrace{\begin{pmatrix} H_0(x) \\ H_1(x) \\ H_2(x) \\ \vdots \end{pmatrix}}_{H} = \underbrace{\begin{pmatrix} 0 & \frac{1}{2} \\ 1 & 0 & \frac{1}{2} \\ & 2 & 0 & \frac{1}{2} \\ & & 3 & 0 & \ddots \\ & & & \ddots & \ddots \end{pmatrix}}_{H} \begin{pmatrix} H_0(x) \\ H_1(x) \\ H_2(x) \\ \vdots \end{pmatrix}$$

Let $\boldsymbol{H} = D^{-1}\boldsymbol{Q}$, where $D = \operatorname{diag}(d_0, d_1, \ldots)$ and $\boldsymbol{Q} = (q_0(x), q_1(x), \ldots)^{\top}$ denotes the (infinite) vector of orthonormal Hermite polynomials. We want to conjugate by D so that DJD^{-1} is symmetric because $x\boldsymbol{Q} = DJD^{-1}\boldsymbol{Q}$. We have

$$DJD^{-1} = \begin{pmatrix} d_0 & & & \\ & d_1 & & \\ & & d_2 & \\ & & & \ddots \end{pmatrix} J \begin{pmatrix} d_0^{-1} & & & \\ & d_1^{-1} & & \\ & & & d_2^{-1} & \\ & & & & \ddots \end{pmatrix} = \begin{pmatrix} 0 & \frac{d_0}{2d_1} & & \\ \frac{d_1}{d_0} & 0 & \frac{d_1}{2d_2} & & \\ & \frac{2d_2}{d_1} & 0 & \frac{d_2}{2d_3} & & \\ & & \frac{3d_3}{d_2} & 0 & \ddots & \\ & & & \ddots & \ddots \end{pmatrix}$$

We require:

$$d_1 d_0^{-1} = \frac{d_0}{2d_1} \Rightarrow d_1^2 = \frac{d_0^2}{2}$$

$$2d_2 d_1^{-1} = \frac{d_1}{2d_2} \Rightarrow d_2^2 = \frac{d_1^2}{4} = \frac{d_0^2}{8} = \frac{d_0^2}{2^2 2!}$$

$$3d_3 d_2^{-1} = \frac{d_2}{2d_3} \Rightarrow d_3^2 = \frac{d_2^2}{3 \times 2} = \frac{d_0^2}{2^3 3!}$$

$$\vdots$$

$$d_n^2 = \frac{d_0^2}{2^n n!}.$$

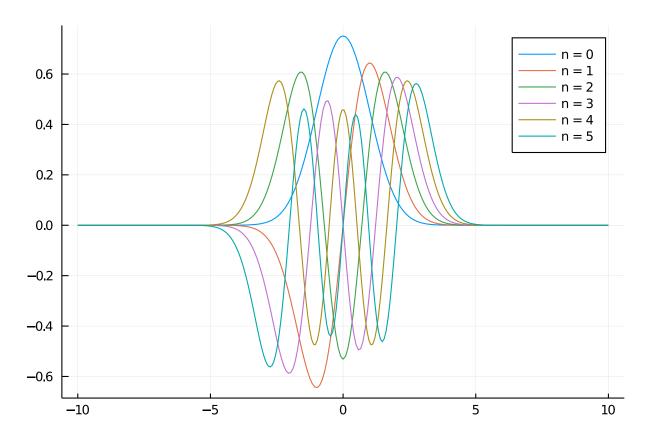
To determine d_0 , note that $q_0 = d_0 H_0 = d_0$ and thus

$$\langle q_0, q_0 \rangle_{\mathrm{H}} = d_0^2 \int_{-\infty}^{\infty} e^{-x^2} dx = d_0^2 \sqrt{\pi} = 1.$$

Hence the orthonormal eigenfunctions in $L^2_{\mathbb{R}}$ with respect to the weight w=1 are

$$\psi_n(x) = q_n(x)e^{-x^2/2} = d_n H_n(x)e^{-x^2/2} = \frac{H_n(x)e^{-x^2/2}}{\sqrt{\sqrt{\pi}2^n n!}}$$

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\begin{array}{lll} p = & plot() \\ & for \ n = 0.5 \\ & H = & Fun(Hermite(), \ [zeros(n);1]) \\ & \psi = & Fun(x \rightarrow H(x)exp(-x^2/2), \ -10.0 \ .. \ 10.0)/sqrt(sqrt(\pi)*2^n*factorial(1.0n)) \\ & plot!(\psi; \ label="n = $n") \\ & end \\ & p \end{array}
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It's convention to shift them by the eigenvalue:

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\begin{array}{lll} p = & plot(pad(Fun(x \rightarrow x^2, -10 ... 10), 100); \ ylims=(0,25)) \\ & for \ n = 0:10 \\ & H = & Fun(Hermite(), \ [zeros(n);1]) \\ & \psi = & Fun(x \rightarrow H(x)exp(-x^2/2), -10.0 ... 10.0)/sqrt(sqrt(\pi)*2^n*factorial(1.0n)) \\ & plot!(\psi + 2n+1; \ label="n = $n") \\ & end \\ & p \end{array}
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