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

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

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)

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

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:

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

if $m < n$.

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

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

∎

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

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

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

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

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:

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

1.10999999999997

We get nonsense when trying to approximate $\text{sech}(x)$ by a degree 50 polynomial:

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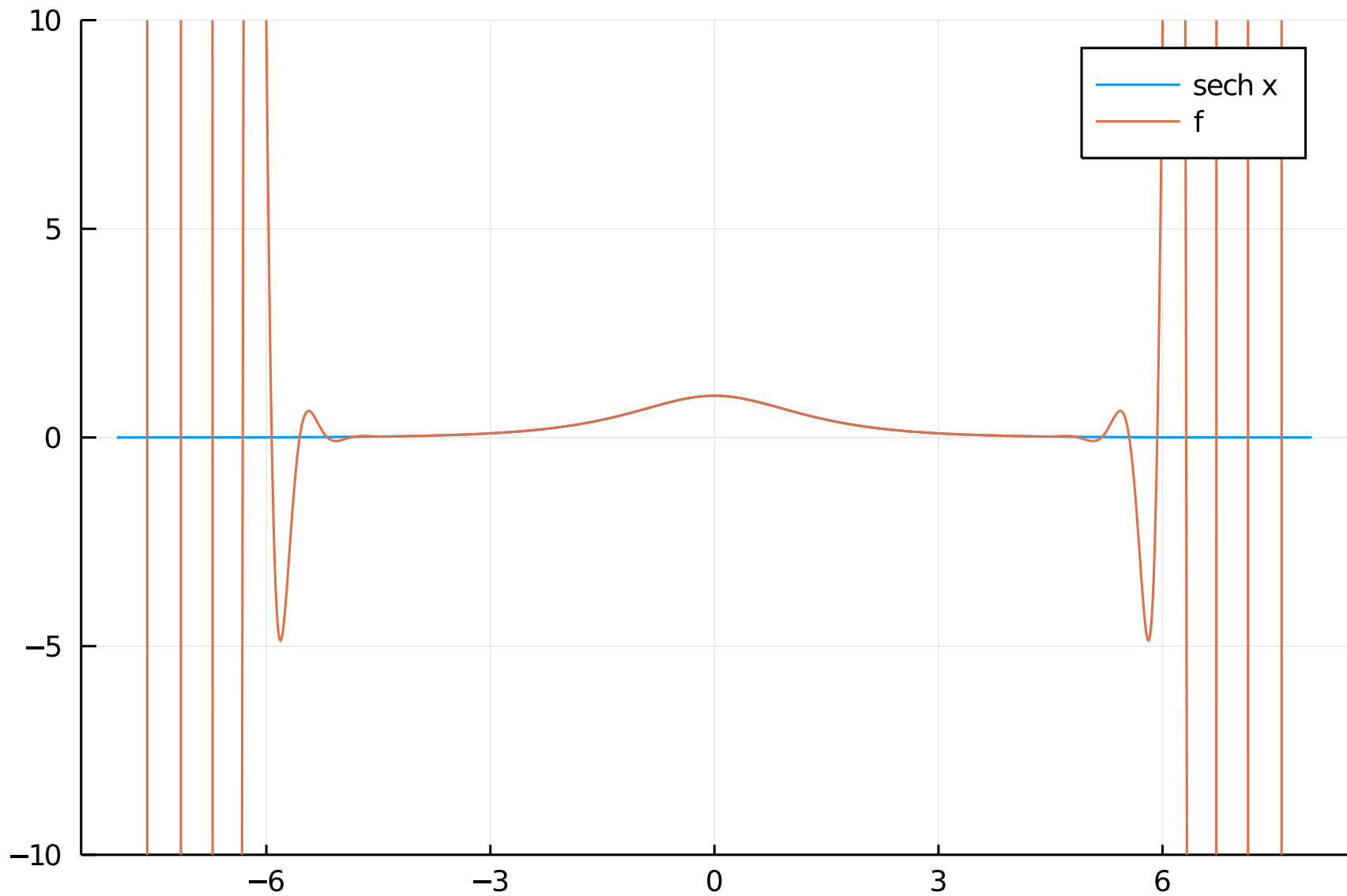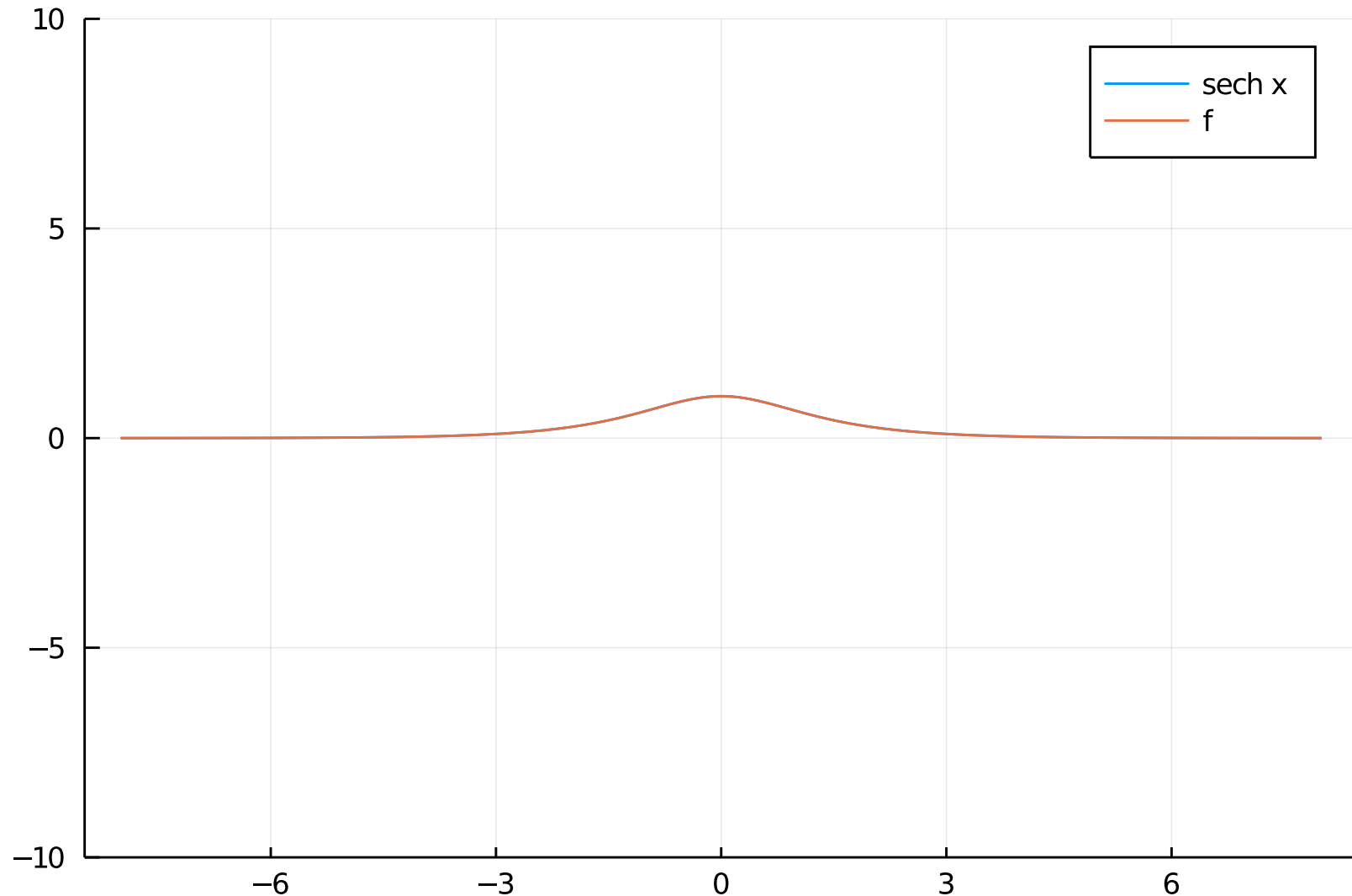
Incorporating the weight $\sqrt{w(x)} = e^{-x^2/2}$ works:

```
f = Fun(x -> sech(x), GaussWeight(Hermite(),1/2),101)

plot(xx, sech.(xx); ylims=(-10,10), label="sech x")
plot!(xx, f.(xx); label="f")
```
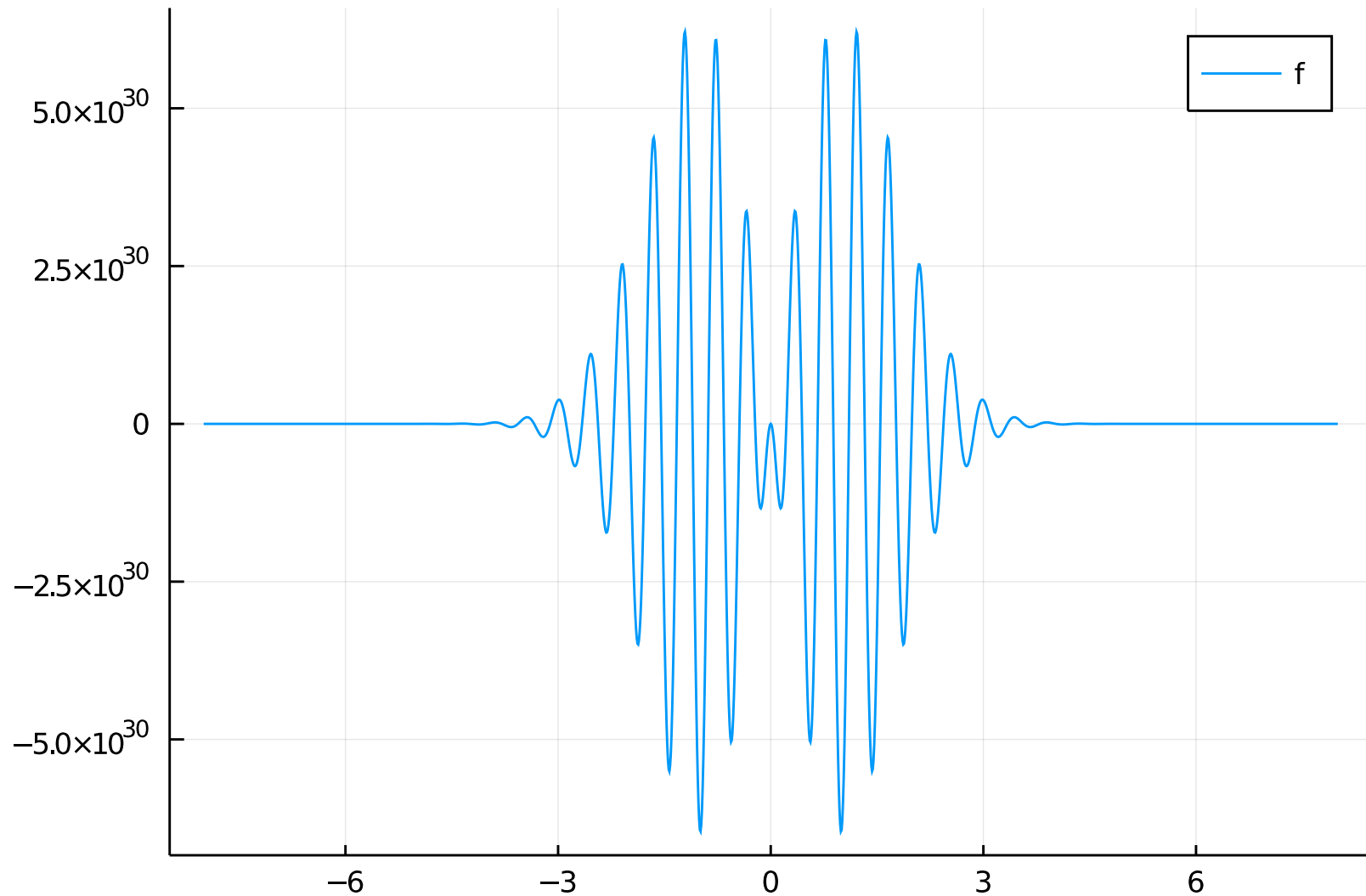
Weighted by $w(x) = e^{-x^2}$ breaks again:

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

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

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

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

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

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

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 Sturm–Liouville operator for Hermite polynomials:

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

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

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

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

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

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

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

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

$$\psi_n(x) = q_n(x) \mathrm{e}^{-x^2/2} = d_n H_n(x) \mathrm{e}^{-x^2/2} = \frac{H_n(x) \mathrm{e}^{-x^2/2}}{\sqrt{\sqrt{\pi} 2^n n!}}$$
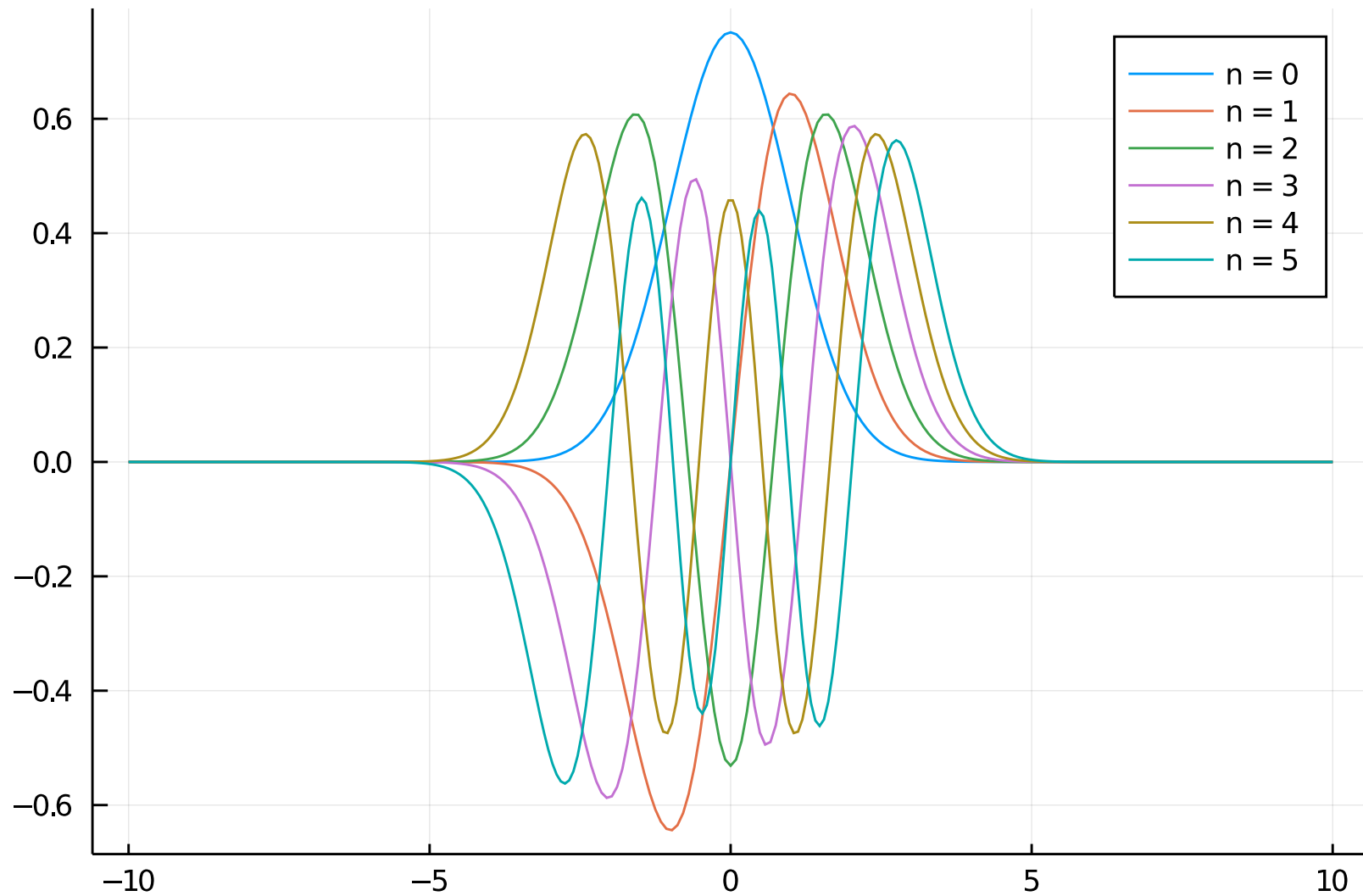
```
p = plot()
for n = 0:5
    H = Fun(Hermite(), [zeros(n);1])
```

```
    ψ = Fun(x -> H(x)exp(-x^2/2), -10.0 ..
10.0)/sqrt(sqrt(π)*2^n*factorial(1.0n))
    plot!(ψ; label="n = $n")
end
p
```

It's convention to shift them by the eigenvalue:

```julia
p = plot(pad(Fun(x -> x^2, -10 .. 10), 100); ylims=(0,25))
for n = 0:10
    H = Fun(Hermite(), [zeros(n);1])
    ψ = Fun(x -> H(x)exp(-x^2/2), -10.0 ..
10.0)/sqrt(sqrt(π)*2^n*factorial(1.0n))
    plot!(ψ + 2n+1; label="n = $n")
end
```