1 Lecture 21: Orthogonal polynomials and singular integrals

This lecture we do the following:

- 1. Cauchy transforms of weighted orthogonal polynomials
 - Three-term recurrence and calculation
 - Hilbert transform of weighted orthogonal polynomials
 - Hilbert transform of weighted Chebyshev polynomials
- 2. Log transform of weighted classical orthogonal polynomials

1.1 Cauchy transforms of orthogonal polynomials

Given a family of orthogonal polynomials $p_k(x)$ with respect to the weight w(x) on (a, b), we always know it satisfies a three-term recurrence:

$$xp_0(x) = a_0p_0(x) + b_0p_1(x)$$

$$xp_k(x) = c_kp_{k-1}(x) + a_kp_k(x) + b_kp_{k+1}(x)$$

Consider now the Cauchy transform of the weighted orthogonal polynomial:

$$C_k(z) := \mathcal{C}_{[a,b]}[p_k w](z) = \frac{1}{2\pi i} \int_a^b \frac{p_k(x)w(x)}{x - z} dx$$

Theorem (Three-term recurrence Cauchy transform of weighted OPs) $C_k(z)$ satisfies the same recurrence relationship as $p_k(x)$ for k = 1, 2, ...:

$$zC_0(z) = a_0C_0(z) + b_0C_1(z) - \frac{1}{2\pi i} \int_a^b w(x) dx$$

$$zC_k(z) = c_kC_{k-1}(z) + a_kC_k(z) + b_kC_{k+1}(z)$$

Proof

$$zC_k(z) = \frac{1}{2\pi i} \int_a^b \frac{zp_k(x)w(x)}{x - z} dx = \frac{1}{2\pi i} \int_a^b \frac{(z - x)p_k(x)w(x)}{x - z} dx + \int_a^b \frac{xp_k(x)w(x)}{x - z} dx$$

$$= -\frac{1}{2\pi i} \int_a^b p_k(x)w(x) dx + \int_a^b \frac{(c_k p_{k-1}(x) + a_k p_k(x) + b_k p_{k+1}(x))w(x)}{x - z} dx$$

$$= -\frac{1}{2\pi i} \int_a^b p_k(x)w(x) dx + c_k C_{k-1}(z) + a_k C_k(z) + b_k C_{k+1}(z)$$

when k > 0, the integral term disappears.

This gives a convenient way to calculate the Cauchy transforms: if we know $C_0(z) = \mathcal{C}w(z)$ and $\int_a^b w(x) dx$, solve the lower triangular system:

$$\begin{pmatrix} 1 & & & & & \\ a_0 - z & b_0 & & & & \\ c_1 & a_1 - z & b_1 & & & \\ & c_2 & a_2 - z & b_2 & & \\ & & c_3 & a_3 - z & \ddots \\ & & & \ddots & \ddots \end{pmatrix} \begin{pmatrix} C_0(z) \\ C_1(z) \\ C_2(z) \\ C_3(z) \\ \vdots \end{pmatrix} = \begin{pmatrix} C_0(z) \\ \frac{1}{2\pi i} \int_a^b w(x) dx \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

Example (Chebyshev Cauchy transform)

Consider the Chebyshev case $w(x) = \frac{1}{\sqrt{1-x^2}}$, which satisfies $\int_{-1}^1 w(x) dx = \pi$. Recall that

$$C_0(z) = \mathcal{C}w(z) = \frac{\mathrm{i}}{2\sqrt{z-1}\sqrt{z+1}}$$

Further, we have

$$xT_0(x) = T_1(x)$$
$$xT_k(x) = \frac{T_{k-1}(x)}{2} + \frac{T_{k+1}(x)}{2}$$

hence

$$zC_0(z) = C_1(z) - \frac{1}{2i}$$
$$zC_k(z) = \frac{C_{k-1}(z)}{2} + \frac{C_{k+1}(z)}{2}.$$

In other words, we want to solve

$$\begin{pmatrix} 1 & & & & \\ -z & 1 & & & \\ 1/2 & -z & 1/2 & & \\ & 1/2 & -z & 1/2 & \\ & & 1/2 & -z & \ddots \\ & & & \ddots & \ddots \end{pmatrix} \begin{pmatrix} C_0(z) \\ C_1(z) \\ C_2(z) \\ C_3(z) \\ \vdots \end{pmatrix} = \begin{pmatrix} \frac{\mathrm{i}}{2\sqrt{z-1}\sqrt{z+1}} \\ \frac{1}{2\mathrm{i}} \\ 0 \\ 0 \\ \vdots \end{pmatrix}$$

with forward substitution.

```
using ApproxFun, SingularIntegralEquations, LinearAlgebra, Plots, ComplexPhasePortrait
x = Fun()
w = 1/sqrt(1-x^2)
z = 0.1+0.1im

n = 10
L = zeros(ComplexF64,n,n)
L[1,1] = 1
```

```
L[2,1] = -z

L[2,2] = 1

for k=3:n

    L[k,k-1] = -z

    L[k,k-2] = L[k,k] = 1/2

end

C = L \ [ im/(2sqrt(z-1)sqrt(z+1)); 1/(2im); zeros(n-2)]

T_5 = Fun(Chebyshev(), [zeros(5);1])

cauchy(T_5*w,z) , C[6]

(0.14734333381379638 - 0.26445831594251407im, 0.14734333381379644 - 0.26445

83159425141im)
```

Warning This fails for large n or large z:

```
x = Fun()
w = 1/sqrt(1-x^2)
z = 5+6im

n = 100
L = zeros(ComplexF64,n,n)
L[1,1] = 1
L[2,1] = -z
L[2,2] = 1
for k=3:n
        L[k,k-1] = -z
        L[k,k-2] = L[k,k] = 1/2
end

C = L \ [ im/(2sqrt(z-1)sqrt(z+1)); 1/(2im); zeros(n-2)]
T_2_0 = Fun(Chebyshev(), [zeros(20);1])
C[21], cauchy(T_2_0*w, z)
(-880764.1597963147 - 1.1245461444433576e6im, 0.0 + 8.834874115176436e-18im)
```

Forward substitution is an unstable algorithm for calculating the $C_k(z)$ because the general solution of the linear recurrence (or linear difference equation) satisfied by the $C_k(z)$ is a superposition (or linear combination) of an exponentially growing solution and an exponentially decaying solution. For large k, the $C_k(z)$ calculated via forward recurrence will pick up the exponentially growing solution because of rounding errors. The computed values of the $C_k(z)$ will then grow exponentially for large k while the actual values the $C_k(z)$ are bounded. Instead, a stable algorithm (e.g., Miller's algorithm or Olver's algorithm) is used to compute the $C_k(z)$. A simple way to compute the C_k more stably is to drop the first row of the recurrence:

```
L[2:end,1:end-1]
C = L[2:end,1:end-1]\ [1/(2im); zeros(n-2)]
C[6]- cauchy(T_5*w, z)
-3.072062106493749e-17 + 4.368604850944066e-17im
```

This algorithm also becomes unstable for large z. For large z, we can compute the C_k using

$$C_k(z) = \frac{1}{2\pi i} \int_a^b \frac{p_k(x)w(x)}{x - z} dx = -\frac{1}{2\pi i z} \int_a^b p_k(x)w(x) \sum_{n=0}^{\infty} \left(\frac{x}{z}\right)^n dx = -\frac{1}{2\pi i} \sum_{n=k}^{\infty} \frac{\mu_n}{z^{n+1}},$$

where μ_n is the *n*-th moment of the OP p_k :

$$\mu_n = \int_a^b x^n p_k(x) w(x) \mathrm{d}x.$$

1.2 Hilbert transform of weighted orthogonal polynomials

Now consider the Hilbert transform of weighted orthogonal polynomials:

$$H_k(x) = \mathcal{H}_{(a,b)}[p_k w](x) = \frac{1}{\pi} \int_a^b \frac{p_k(t)w(t)}{x - t} dt$$

Just like Cauchy transforms, the Hilbert transforms have

Corollary (Hilbert transform recurrence)

$$xH_0(x) = a_0H_0(x) + b_0H_1(x) + \frac{1}{\pi} \int_a^b w(x)dx$$

$$xH_k(x) = c_kH_{k-1}(x) + a_kH_k(x) + b_kH_{k+1}(x)$$

Proof Recall

$$C^+f(x) + C^-f(x) = i\mathcal{H}f(x)$$

Therefore, we have

$$C_k^+(x) + C_k^-(x) = i\mathcal{H}[wp_k](x)$$

hence we have

$$xH_0(x) = -ix(C_0^+(x) + C_0^-(x)) = -i\left[a_0(C_0^+(x) + C_0^-(x)) + b_0(C_1^+(x) + C_1^-(x)) - \frac{1}{\pi i} \int_a^b w(x) dx\right]$$
$$= a_0H_0(x) + b_0H_1(x) + \frac{1}{\pi} \int_a^b w(x) dx$$

Other k follows by a similar argument.

1.2.1 Example 1: weighted Chebyshev T

For

$$H_k(x) := \mathcal{H}[T_k/\sqrt{1-x^2}](x) = \frac{1}{\pi} \int_{-1}^1 \frac{T_k(t)}{(x-t)\sqrt{1-t^2}} dt$$

The recurrence gives us

$$xH_0(x) = H_1(x) + 1$$
$$xH_k(x) = \frac{H_{k-1}(x)}{2} + \frac{H_{k+1}(x)}{2}$$

In this case, we have $H_0(x) = \mathcal{H}[w](x) = 0$. Therefore, we can rewrite this recurrence as

$$H_1(x) = -1$$

$$xH_1(x) = \frac{H_2(x)}{2}$$

$$xH_k(x) = \frac{H_{k-1}(x)}{2} + \frac{H_{k+1}(x)}{2}$$

This is precisely the three-term recurrence satisfied by $-U_{k-1}$! We therefore have

$$H_k(x) = -U_{k-1}(x)$$

This gives a very easy way to compute Hilbert transforms: if

$$f(x) = \sum_{k=0}^{\infty} f_k T_k(x)$$

then

$$\mathcal{H}\left[\frac{f}{\sqrt{1-\diamondsuit^2}}\right](x) = -\sum_{k=0}^{\infty} f_{k+1}U_k(x)$$

```
x = 0.1
T = Fun(Chebyshev(),[zeros(n);1])
H(f,x) = -hilbert(f,x) # Fix normalisation
H(w*T,x) , -Fun(Ultraspherical(1), [zeros(n-1);1])(x)
```

1.2.2 Example 2: weighted Chebyshev U

(-0.5608031061203765, -0.5608031061203765)

For

$$H_k(x) := \mathcal{H}[U_k\sqrt{1-x^2}](x) = \frac{1}{\pi} \int_{-1}^1 \frac{U_k(t)\sqrt{1-t^2}}{x-t} dt$$

The recurrence gives us

$$xH_0(x) = \frac{H_1(x)}{2} + 1/2$$
$$xH_k(x) = \frac{H_{k-1}(x)}{2} + \frac{H_k(x)}{2}$$

In this case, we have $H_0(x) = \mathcal{H}[w](x) = x$. Therefore, we can rewrite this recurrence as

$$H_{-1}(x) := 1$$

$$xH_{-1}(x) = H_0(x)$$

$$xH_0(x) = \frac{H_{-1}(x)}{2} + \frac{H_1(x)}{2}$$

$$xH_k(x) = \frac{H_{k-1}(x)}{2} + \frac{H_{k+1}(x)}{2}$$

This is precisely the three-term recurrence satisfied by T_{k+1} ! We therefore have

$$H_k(x) = T_{k+1}(x)$$

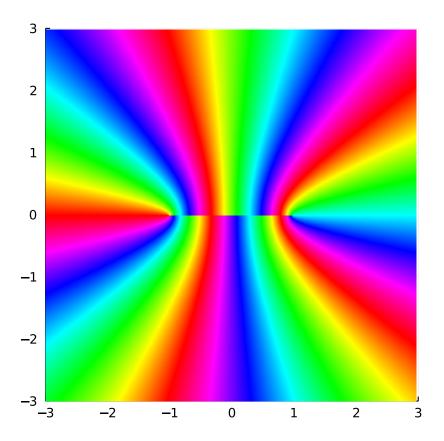
1.3 Log transforms of weighted orthogonal polynomials

Now consider $\frac{1}{\pi} \int_a^b p_k(x) w(x) \log |z - x| dx$, which we write in terms of the real part of

$$M_k(z) = M[p_k w](z) = \frac{1}{\pi} \int_a^b p_k(x) w(x) \log(z - x) dx$$
$$= \frac{\log(z - a)}{\pi} \int_a^b p_k(x) w(x) dx + 2i\mathcal{C}_{[a,b]} F(z)$$

where $F(x) = \int_x^1 p_k(t)w(t)dt$. For k > 0 we have $\int_a^b p_k(x)w(x)dx = 0$ due to orthogonality, and hence we actually have no branch cut on $(-\infty, a)$:

```
x = Fun()
T_5 = Fun(Chebyshev(), [zeros(5);1])
w = 1/sqrt(1-x^2)
M_5 = z-> -2im*cauchyintegral(w*T_5, z) # cauchyintegral computes an indefinite integral
of the Cauchy transform
phaseplot(-3..3, -3..3, M_5)
```



1.4 Weighted Chebyshev log transform

For classical orthogonal polynomials we can go a step further and relate the indefinite integrals to other orthogonal polynomials.

For example, recall that

$$\frac{\mathrm{d}}{\mathrm{d}x}[\sqrt{1-x^2}U_n(x)] = -\frac{n+1}{\sqrt{1-x^2}}T_{n+1}(x)$$

in other words,

$$\int_{x}^{1} \frac{T_{k}(t)}{\sqrt{1-t^{2}}} dt = \frac{\sqrt{1-x^{2}}U_{k-1}(x)}{k}$$

Thus for k = 1, 2, ...,

$$M_k(z) = \frac{1}{\pi} \int_{-1}^1 \frac{T_k(x)}{\sqrt{1 - x^2}} \log(z - x) dx = \frac{2i}{k} \mathcal{C}[\sqrt{1 - \diamond^2} U_{k-1}](z)$$

and for k = 0,

$$M_0(z) = 2\log(\sqrt{z-1} + \sqrt{z+1}) - 2\log 2$$

As we saw above, Cauchy transforms of weighted OPs satisfy simple recurrences, and this relationship renders log transforms equally calculable.

```
T_5 = Fun(Chebyshev(), [zeros(5);1])
U_4 = Fun(Ultraspherical(1), [zeros(4);1])
```

```
x = Fun()

M_5 = z - sum(T_5/sqrt(1-x^2) * log(z-x))/\pi

M_5(z), 2im*cauchy(sqrt(1-x^2)*U_4,z)/5
```

(6.576121814966234e-8 - 2.0389718362285198e-7im, 6.576121814966217e-8 - 2.0389718362285196e-7im)

For $z = x \in [-1, 1]$ and k > 0

$$L_{k}(x) = \frac{1}{\pi} \int_{-1}^{1} \frac{T_{k}(t) \log |t - x|}{\sqrt{1 - t^{2}}} dt$$

$$= \Re M_{k}^{+}(x)$$

$$= -\frac{2}{k} \Im \mathcal{C}^{+} [\sqrt{1 - \diamond^{2}} U_{k-1}](x)$$

$$= -\frac{\mathcal{H}[\sqrt{1 - \diamond^{2}} U_{k-1}](x)}{k}$$

$$= -\frac{T_{k}(x)}{k}$$

and

$$L_0(x) = \Re M_0^+(x) = -\log 2$$

$$x_0 = 1/sqrt(2)$$

 $L_5(x) = logkernel(w*T_5,x)$
 $L_5(x_0), -T_5(x_0)/5$

(0.14142135623730945, 0.14142135623730942)