

Polynomials orthogonal with respect to densely oscillating and exponentially decaying weight functions

Walter Gautschi

wxg@cs.purdue.edu

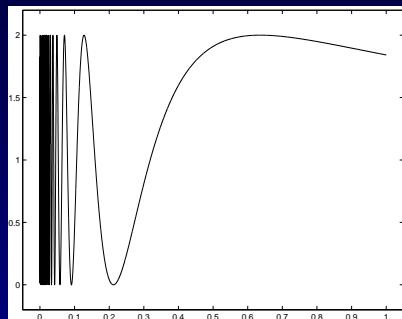
Purdue University

Web Site

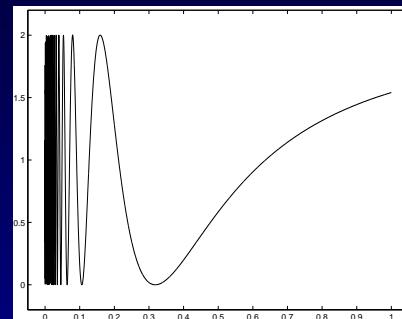
[http://www.cs.purdue.edu/
archives/2002/wxg/codes](http://www.cs.purdue.edu/archives/2002/wxg/codes)

A gallery of weight functions

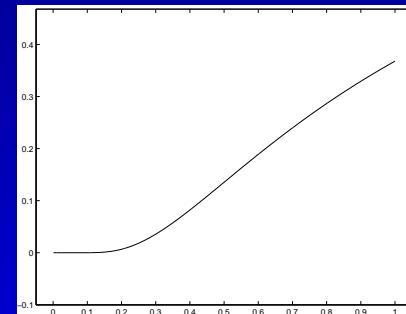
$$1 + \sin(1/t)$$



$$1 + \cos(1/t)$$

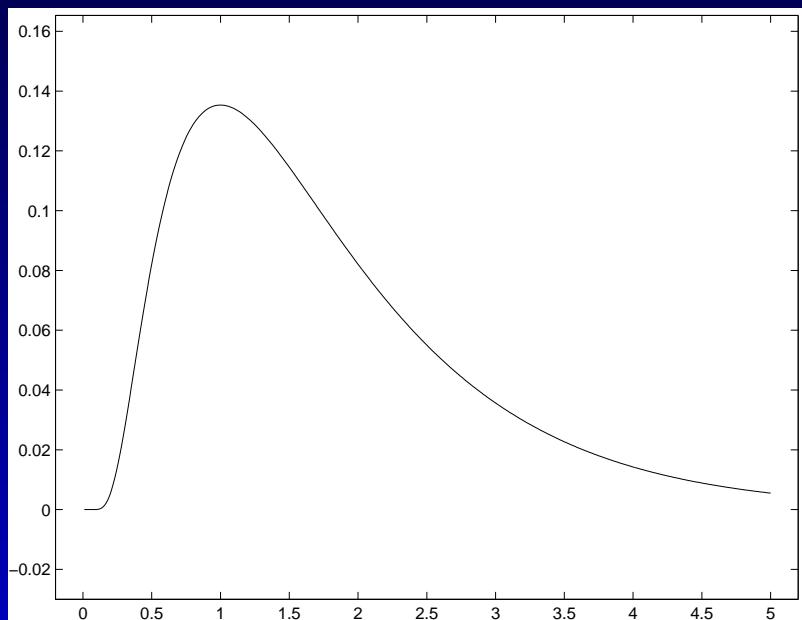


$$\exp(-1/t)$$

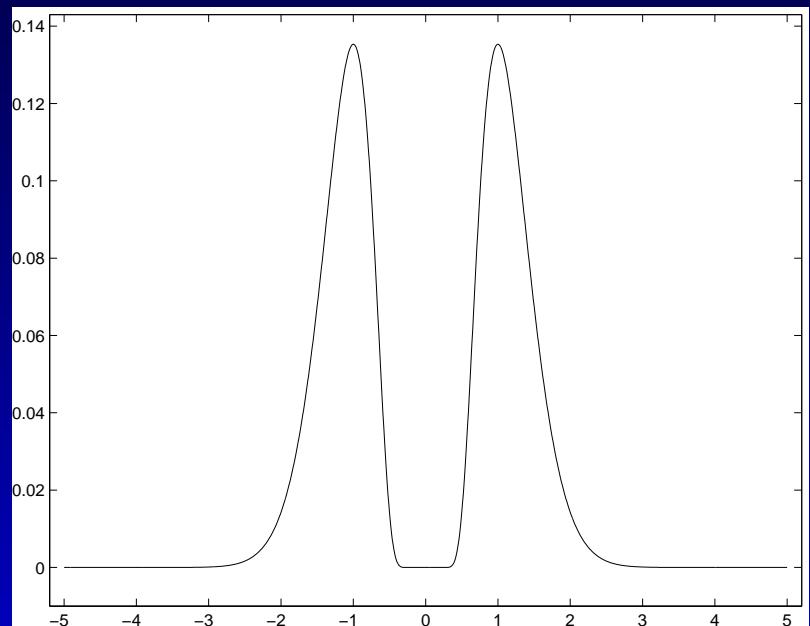


gallery (cont')

$$\exp(-1/t - t)$$



$$\exp(-1/t^2 - t^2)$$



Moment-based approach

The first $2n$ moments

$$\mu_k = \int_a^b t^k w(t) dt, \quad k = 0, 1, \dots, 2n - 1,$$

of a weight function w on $[a, b]$ determine uniquely the first n recurrence coefficients $\alpha_k, \beta_k, k = 0, 1, \dots, n - 1$ in the three-term recurrence relation for the orthogonal polynomials $\pi_k(\cdot; w)$,

$$\pi_{k+1}(t) = (t - \alpha_k)\pi_k(t) - \beta_k\pi_{k-1}(t), \\ k = 0, 1, \dots, n - 1,$$

$$\pi_{-1}(t) = 0, \quad \pi_0(t) = 1$$

Moment map

$$\text{mom} = [\mu_k]_{k=0}^{2n-1} \mapsto \mathbf{ab} = [\alpha_k, \beta_k]_{k=0}^{n-1}$$

Moment map

$$\text{mom} = [\mu_k]_{k=0}^{2n-1} \mapsto \text{ab} = [\alpha_k, \beta_k]_{k=0}^{n-1}$$

Implementation

Chebyshev algorithm (1859)

Moment map

$$\text{mom} = [\mu_k]_{k=0}^{2n-1} \mapsto \mathbf{ab} = [\alpha_k, \beta_k]_{k=0}^{n-1}$$

Implementation

Chebyshev algorithm (1859)

Matlab

```
ab=chebyshev(N,mom)
```

Moment map

$$\text{mom} = [\mu_k]_{k=0}^{2n-1} \mapsto \mathbf{ab} = [\alpha_k, \beta_k]_{k=0}^{n-1}$$

Implementation

Chebyshev algorithm (1859)

Matlab

`ab=chebyshev(N, mom)`

Ill-conditioned!

Moment map

$$\text{mom} = [\mu_k]_{k=0}^{2n-1} \mapsto \mathbf{ab} = [\alpha_k, \beta_k]_{k=0}^{n-1}$$

Implementation

Chebyshev algorithm (1859)

Matlab

`ab=chebyshev(N, mom)`

Ill-conditioned!

Remedy: symbolic Matlab

`ab=schebyshev(dig, N, mom)`

The weight function $w(t) = 1 + \sin(1/t)$ **on** $[0, 1]$
moments

let

$$\mu_k^0 = \int_0^1 t^k \sin(1/t) dt = \int_1^\infty t^{-(k+2)} \sin t dt$$

then

$$\mu_{k+1}^0 = \frac{1}{k+2} \left[\frac{1}{k+1} (\cos 1 - \mu_{k-1}^0) + \sin 1 \right]$$

where

$$\mu_{-1}^0 = \frac{\pi}{2} - \text{Si}(1), \quad \mu_0^0 = \sin 1 - \text{Ci}(1)$$

The weight function $w(t) = 1 + \sin(1/t)$ **on** $[0, 1]$
moments

let

$$\mu_k^0 = \int_0^1 t^k \sin(1/t) dt = \int_1^\infty t^{-(k+2)} \sin t dt$$

then

$$\mu_{k+1}^0 = \frac{1}{k+2} \left[\frac{1}{k+1} (\cos 1 - \mu_{k-1}^0) + \sin 1 \right]$$

where

$$\mu_{-1}^0 = \frac{\pi}{2} - \text{Si}(1), \quad \mu_0^0 = \sin 1 - \text{Ci}(1)$$

thus

$$\mu_k = \int_0^1 t^k [1 + \sin(1/t)] dt = \frac{1}{k+1} + \mu_k^0$$

Matlab routine

```
% SR_SINO Symbolic/variable-precision recurrence
%   coefficients for the weight function w(x)=
%   1+sin(1/x) on [0,1]
%
%  
syms mom ab  
digits(d); dig=d;  
mom(1)='sin(1)-Ci(1)';  
mom(2)='(Si(1)-pi/2+sin(1)+cos(1))/2';  
for k=3:2*N  
    mom(k)=(( 'cos(1)', -mom(k-2)) / (k-1) + 'sin(1)' ) / k;  
end  
for k=1:2*N  
    mom(k)=mom(k)+1/k;  
end  
ab=schebyshew(dig,N,mom);
```

output (N=40, d=16)

```
ab =  
[.5841029561609566, 1.504067061906928]  
[.4634474607770499, .7094822535096882e-1]  
[.4977629714178322, .7892077774694954e-1]  
[.5356590088623750, .5547885019105795e-1]  
[.4669144430825117, .6259489484316939e-1]  
[.4951204560106238, .7262419731845188e-1]  
    ...    ...    ...  
[.5109646577717308, .5919672749794708e-1]  
[.4919346767523436, .6218164541801303e-1]  
[.4973070307106440, .6494413841533854e-1]  
[.5100281141083978, .6093638294208964e-1]
```

Example 1 $\int_0^1 \tan((\frac{1}{2}\pi - \delta)t) \sin(1/t) dt$

Example 1 $\int_0^1 \tan((\frac{1}{2}\pi - \delta)t) \sin(1/t) dt$

$$\int_0^1 f(t) \sin(1/t) dt = \int_0^1 f(t)[1 + \sin(1/t)] dt - \int_0^1 f(t) dt$$

Example 1 $\int_0^1 \tan((\frac{1}{2}\pi - \delta)t) \sin(1/t) dt$

$$\int_0^1 f(t) \sin(1/t) dt = \int_0^1 f(t)[1 + \sin(1/t)] dt - \int_0^1 f(t) dt$$

```
% INTSINO Example 1
%
load -ascii absin0;
ab=absin0; abl=r_jacobi01(40);
for n=4:4:35
    xwl=gauss(n,abl); xw=gauss(n,ab);
    intl=sum(xwl(:,2).*tan((pi/2-delta)...
        .*xwl(:,1)));
    ints=sum(xw(:,2).*tan((pi/2-delta)...
        .*xw(:,1)));
    int=ints-intl
end
```

output ($\delta = .1$)

```
>> intsin0  
  
1.2716655036125e+00  
1.2957389942560e+00  
1.2961790099686e+00  
1.2961860624657e+00  
1.2961861691603e+00  
1.2961861708344e+00  
1.2961861708631e+00  
1.2961861708636e+00  
1.2961861708636e+00
```

output ($\delta = .1$)

```
>> intsin0
1.2716655036125e+00
1.2957389942560e+00
1.2961790099686e+00
1.2961860624657e+00
1.2961861691603e+00
1.2961861708344e+00
1.2961861708631e+00
1.2961861708636e+00
1.2961861708636e+00
```

1000-point Gauss-Legendre
on [0,1] yields 6 correct digits

The weight function $w(t) = e^{-1/t}$ **on** $[0, 1]$

moments

$$\mu_k = \int_0^1 t^k e^{-1/t} dt = \int_1^\infty t^{-(k+2)} e^{-t} dt = E_{k+2}(1)$$

recurrence

$$\mu_{k+1} = \frac{1}{k+2} (e^{-1} - \mu_k), \quad \mu_0 = E_2(1)$$

The weight function $w(t) = e^{-1/t}$ on $[0, 1]$

moments

$$\mu_k = \int_0^1 t^k e^{-1/t} dt = \int_1^\infty t^{-(k+2)} e^{-t} dt = E_{k+2}(1)$$

recurrence

$$\mu_{k+1} = \frac{1}{k+2} (e^{-1} - \mu_k), \quad \mu_0 = E_2(1)$$

Matlab routine

```
% SR_EXPO Symbolic/variable-precision recurrence coefficients
% for the weight function w(x)=exp(-1/x) on [0,1]
%
syms mom ab
digits(d); dig=d;
mom(1)=vpa('Ei(2,1)',d);
for k=2:2*N
    mom(k)=('exp(-1)',-mom(k-1))/k;
end
ab=schebyshев(dig,N,mom);
```

Example 2 $\int_0^1 \ln(1 + t)e^{-1/t} dt$

Example 2 $\int_0^1 \ln(1+t)e^{-1/t} dt$

Gauss quadrature with weight function $e^{-1/t}$

```
>> intexp0
```

n	n-point Gauss
2	8.1262554100479e-02
4	8.1255735149253e-02
6	8.1255733983155e-02
8	8.1255733982820e-02
10	8.1255733982819e-02
12	8.1255733982819e-02

102-point Gauss–Laguerre quadrature of $e^{-1}(1+t)^{-2} \ln(1+(1+t)^{-1})$ yields the same limit value

The weight function $w(t) = e^{-1/t-t}$ **on** $[0, \infty]$
moments

$$\mu_k = \int_0^\infty t^k e^{-(1/t+t)} dt = 2K_{k+1}(2)$$

recurrence

$$\mu_{k+1} = (k+1)\mu_k + \mu_{k-1}, \quad \mu_{-1} = 2K_0(2), \quad \mu_0 = 2K_1(2)$$

The weight function $w(t) = e^{-1/t-t}$ on $[0, \infty]$
moments

$$\mu_k = \int_0^\infty t^k e^{-(1/t+t)} dt = 2K_{k+1}(2)$$

recurrence

$$\mu_{k+1} = (k+1)\mu_k + \mu_{k-1}, \quad \mu_{-1} = 2K_0(2), \quad \mu_0 = 2K_1(2)$$

Matlab routine

```
% SR_EXPOINF Variable-precision recurrence coefficients
%   for the weight function w(x)=exp(-1/x-x) on [0,inf]
%
syms mom ab
digits(d); dig=d;
mom(1)=vpa('2*BesselK(1,2)',dig);
mom(2)=mom(1)+vpa('2*BesselK(0,2)',dig);
for k=3:2*N
    mom(k)=(k-1)*mom(k-1)+mom(k-2);
end
ab=schebyshев(dig,N,mom);
```

output (N=40, d=16)

```
ab =  
[ 1.814307758763789, .2797317636330449]  
[ 3.647885050815283, 1.336902874017094]  
[ 5.563608408242503, 4.576187502809998]  
[ 7.510248881089434, 9.776110045536486]  
[ 9.472385776425876, 16.95364518291704]  
[ 11.44360258233455, 26.11622048172850]  
    ...  
    ...  
[ 73.23900952424970, 1300.294223771922]  
[ 75.23687505008481, 1373.374066736622]  
[ 77.23481475899122, 1448.453190623454]  
[ 79.23282425676095, 1525.531620678047]
```

Example 3 $\int_0^\infty J_0(t)e^{-1/t-t}dt$

Example 3 $\int_0^\infty J_0(t)e^{-1/t-t}dt$
Gauss quadrature with weight function $e^{-1/t-t}$

```
>> intexp0inf
```

n	n-point Gauss
4	1.1162402700893e-01
8	1.1153340191221e-01
12	1.1153288987809e-01
16	1.1153289176609e-01
20	1.1153289176207e-01
24	1.1153289176207e-01

1000-point Gauss–Laguerre quadrature of
 $J_0(t)e^{-1/t}$ yields eleven correct digits

The weight function $w(t) = e^{-1/t^2 - t^2}$ **on** \mathbb{R}
moments

$$\mu_{2k+1} = 0, \quad \mu_{2k} = 2K_{k+1/2}(2)$$

recurrence

$$\mu_{2k+2} = (k + \frac{1}{2})\mu_{2k} + \mu_{2k-2}, \quad \mu_{-2} = \mu_0 = 2K_{1/2}(2)$$

The weight function $w(t) = e^{-1/t^2-t^2}$ on \mathbb{R}
moments

$$\mu_{2k+1} = 0, \quad \mu_{2k} = 2K_{k+1/2}(2)$$

recurrence

$$\mu_{2k+2} = (k + \frac{1}{2})\mu_{2k} + \mu_{2k-2}, \quad \mu_{-2} = \mu_0 = 2K_{1/2}(2)$$

Matlab routine

```
% SR_EXPMINFPINF Symbolic/variable-precision recurrence
% coefficients for the weight function
% w(x)=exp(-1/x^2-x^2) on [-inf,inf]
%
syms mom ab
digits(d); dig=d;
mom(2:2:2*N)=vpa(0,dig);
mom(1)=vpa('2*BesselK(1/2,2)',dig);
mom(3)=3*mom(1)/2;
for k=3:2*N
    mom(2*k-1)=(k-3/2)*mom(2*k-3)+mom(2*k-5);
end
ab=schebyshrev(dig,N,mom);
```

output (N=40, d=16)

```
ab =  
[0, .2398755439361229]  
[0, 1.500000000000000]  
[0, .6666666666666667]  
[0, 2.58333333333333]  
[0, 1.475806451612903]  
[0, 3.659439450026441]  
... ...  
[0, 16.49963635631090]  
[0, 20.31604933808658]  
[0, 17.46711169129393]  
[0, 21.34281282504185]
```

Example 4 $\int_{-\infty}^{\infty} e^{-t} \cos t e^{-1/t^2 - t^2} dt$

Example 4 $\int_{-\infty}^{\infty} e^{-t} \cos t e^{-1/t^2-t^2} dt$
Gauss quadrature with weight function e^{-1/t^2-t^2}

```
>> intexpminfpinf
```

n	n-point Gauss
2	1.5040374279876e-01
4	1.1308193957943e-01
6	1.1342796227803e-01
8	1.1342695840745e-01
10	1.1342695981592e-01
12	1.1342695981475e-01
14	1.1342695981475e-01

1000-point Gauss–Hermite quadrature of
 $e^{-t-1/t^2} \cos t$ yields eight correct digits

Circle theorem for Gaussian quadrature

(Davis and Rabinowitz, 1961)

Theorem The nodes τ_ν and weights λ_ν of the n -point Gaussian quadrature formula on $[-1, 1]$ satisfy

$$n\lambda_\nu / (\pi w(\tau_\nu)) \sim \sqrt{1 - \tau_\nu^2}, \quad n \rightarrow \infty.$$

Circle theorem for Gaussian quadrature

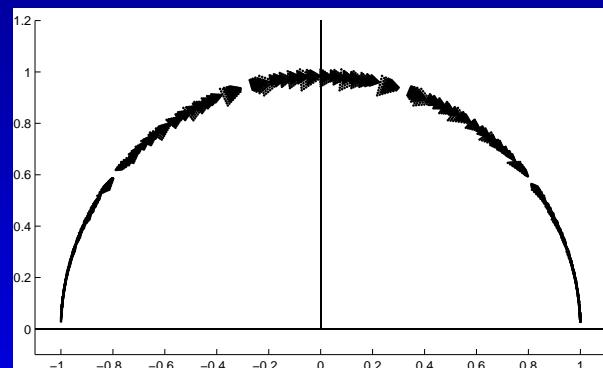
(Davis and Rabinowitz, 1961)

Theorem The nodes τ_ν and weights λ_ν of the n -point Gaussian quadrature formula on $[-1, 1]$ satisfy

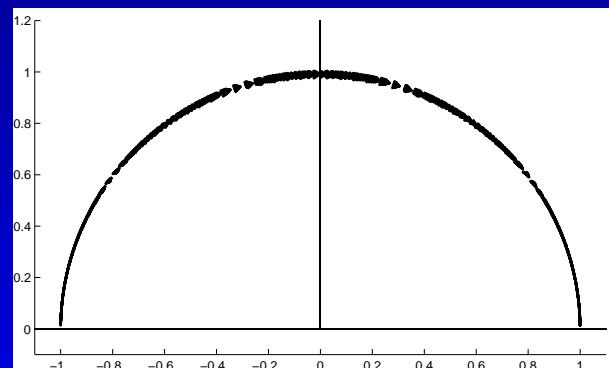
$$n\lambda_\nu / (\pi w(\tau_\nu)) \sim \sqrt{1 - \tau_\nu^2}, \quad n \rightarrow \infty.$$

Example Gauss-Jacobi formula, with Jacobi parameters

$$\alpha, \beta = -.75 : .25 : 1.0, \beta \geq \alpha$$



$$n = 20 : 5 : 40$$



$$n = 60 : 5 : 80$$

Circle theorem for Gaussian quadrature

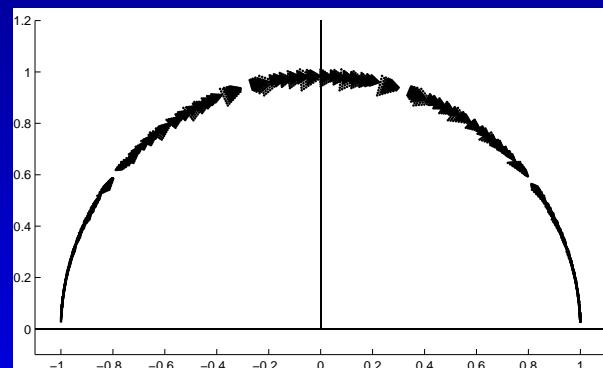
(Davis and Rabinowitz, 1961)

Theorem The nodes τ_ν and weights λ_ν of the n -point Gaussian quadrature formula on $[-1, 1]$ satisfy

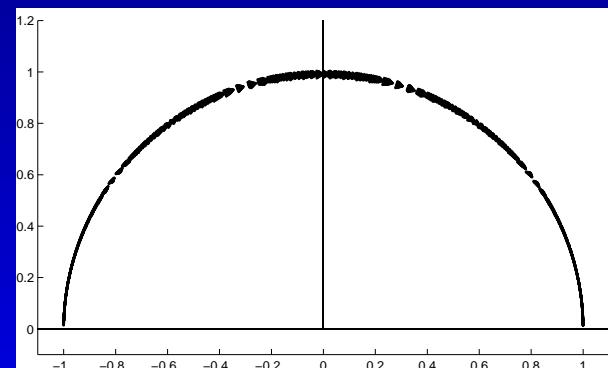
$$n\lambda_\nu / (\pi w(\tau_\nu)) \sim \sqrt{1 - \tau_\nu^2}, \quad n \rightarrow \infty.$$

Example Gauss-Jacobi formula, with Jacobi parameters

$$\alpha, \beta = -.75 : .25 : 1.0, \beta \geq \alpha$$



$$n = 20 : 5 : 40$$

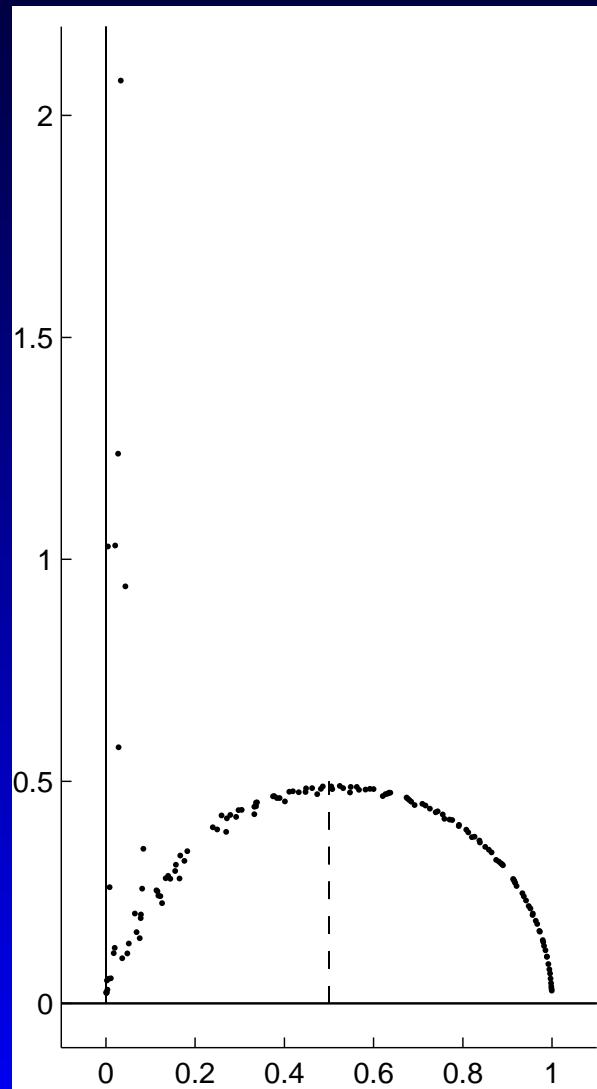


$$n = 60 : 5 : 80$$

Query True for $1 + \sin(1/t)$ or $\exp(-1/t)$?

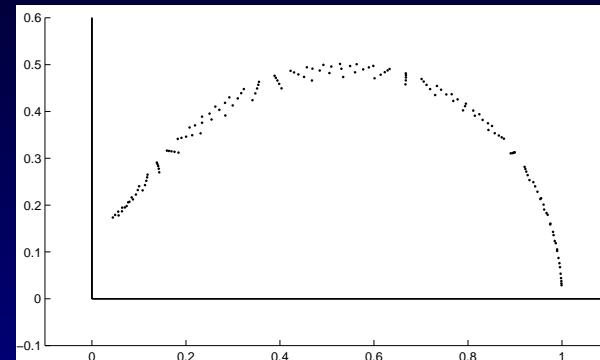
The circle theorem gone berserk

$$w(t) = 1 + \sin(1/t) \text{ on } [0, 1]$$



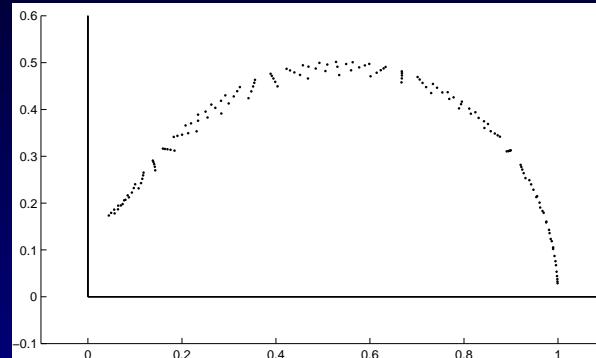
The circle theorem impaired

$$w(t) = \exp(-1/t) \text{ on } [0, 1]$$



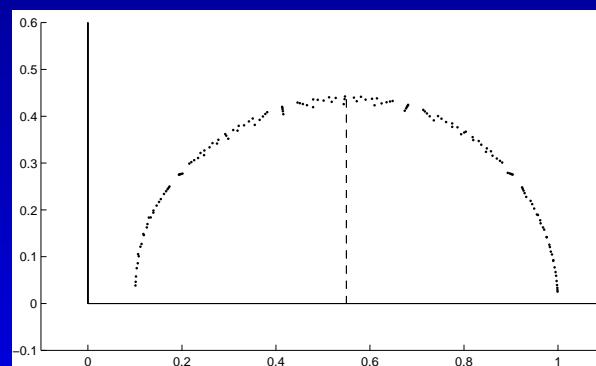
The circle theorem impaired

$$w(t) = \exp(-1/t) \text{ on } [0, 1]$$



...and repaired

$$w(t) = \exp(-1/t) \text{ on } [c, 1], c = .1$$



The circle theorem for weight functions in the Szegö class

Theorem (Nevai, 1979). If $\log w(t)/\sqrt{1-t^2} \in L_1[-1, 1]$ and $1/(\sqrt{1-t^2} w(t)) \in L_1[\Delta]$, where Δ is any compact subinterval of $(-1, 1)$, then

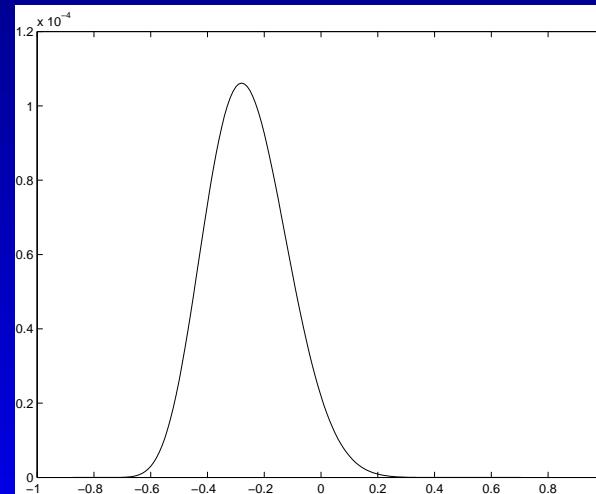
$$n \lambda_\nu / (\pi w(\tau_\nu)) \sim \sqrt{1 - \tau_\nu^2} \text{ on } \Delta, \quad n \rightarrow \infty.$$

The circle theorem for weight functions in the Szegö class

Theorem (Nevai, 1979). If $\log w(t)/\sqrt{1-t^2} \in L_1[-1, 1]$ and $1/(\sqrt{1-t^2} w(t)) \in L_1[\Delta]$, where Δ is any compact subinterval of $(-1, 1)$, then

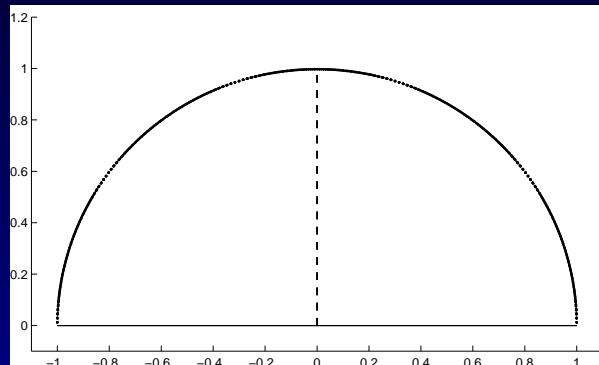
$$n \lambda_\nu / (\pi w(\tau_\nu)) \sim \sqrt{1 - \tau_\nu^2} \text{ on } \Delta, \quad n \rightarrow \infty.$$

Example Pollaczek weight function $w(\cdot; a, b)$, $a \geq |b|$



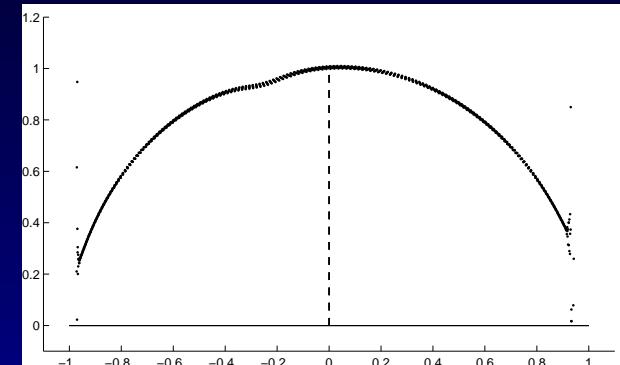
$$a = 4, b = 1$$

“Circle theorem” for Pollaczek weight function



$$a = b = 0$$

$$n = 180 : 5 : 200$$



$$a = 4, b = 1$$