

The Hilbert matrix:

$$H_n = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \cdots & \frac{1}{n} \\ \frac{1}{2} & \frac{1}{3} & \frac{1}{4} & \cdots & \frac{1}{n+1} \\ \frac{1}{3} & \frac{1}{4} & \frac{1}{5} & \cdots & \frac{1}{n+2} \\ \vdots & \vdots & \vdots & \ddots & \\ \frac{1}{n} & \frac{1}{n+1} & \frac{1}{n+2} & \cdots & \frac{1}{2n-1} \end{bmatrix}.$$

is a great example to illustrate ill-conditioning. For a symmetric ill-conditioned matrix, it is useful to examine sensitivities round eigenvectors as they all "point" in different directions, colloquially speaking.

H = #1 (generic function with 1 method)

```
. H = n -> inv.((1:n) .+ (1:n)' .- 1)
```

A = 15×15 Array{Float64,2}:

```
1.0      0.5      ...  0.06666666666666667
0.5      0.3333333333333333  0.0625
0.3333333333333333  0.25      0.058823529411764705
0.25     0.2       0.05555555555555555
0.2      0.16666666666666666  0.05263157894736842
0.16666666666666666  0.14285714285714285 ... 0.05
0.14285714285714285  0.125     0.047619047619047616
⋮
0.1      0.09090909090909091  0.041666666666666664
0.09090909090909091  0.08333333333333333 ... 0.04
0.08333333333333333  0.07692307692307693  0.038461538461538464
0.07692307692307693  0.07142857142857142  0.037037037037037035
0.07142857142857142  0.06666666666666667  0.03571428571428571
0.06666666666666667  0.0625     0.034482758620689655
```

```
. A = H(15)
```

```
. using LinearAlgebra
```

```
Eigen{Float64,Float64,Array{Float64,2},Array{Float64,1}}
```

```
values:
```

```
15-element Array{Float64,1}:
```

```
-9.731674398306353e-18
 2.4471098684428205e-17
 8.417476172219603e-17
 1.3900578237791628e-14
 9.321996824788568e-13
 4.657781739919007e-11
 1.8029597497893784e-9
  :
 2.710853922820439e-5
 0.0004364765944177553
 0.005639834755789189
 0.05721209253338421
 0.4266279570069751
 1.845927746153489
```

```
vectors:
```

```
15×15 Array{Float64,2}:
```

```
 4.027407113534054e-9   -3.8579037815039685e-8   ...   -0.6643596446265618
-6.434951197145829e-7    5.259822492602317e-6   ...   -0.41379493314732985
 2.5683011205110927e-5   -0.00017821662657140948   -0.31263025106057074
-0.00044963564925012576  0.0026140055386670736   -0.25497095328304914
 0.004312625558763815   -0.020536398884193487   -0.21683971746073358
-0.02541271764022335    0.09560128949271934   ...   -0.18940844991079595
 0.09820458417360232   -0.2748896282066456   ...   -0.16856873009438833
  :
-0.5937090230237423     0.028349623470173134   ...   -0.12765836086544508
 0.5096716554877957     0.42890719110460424   ...   -0.11827942474402993
-0.2840155991095513     -0.4892558710585028   ...   -0.11023685308104106
 0.09265662686030433     0.24070759428147576   ...   -0.10325655189669361
-0.013433180119699654   -0.04659932577569513   ...   -0.09713593495171591
 0.0                      0.0                      ...   -0.0917217396947767
```

```
Λ, V = eigen(A)
```

```
x = Float64[4.02741e-9, -6.43495e-7, 2.5683e-5, -0.000449636, 0.00431263, -0.0254
```

```
x = V[:,1]
```

```
Δx = Float64[-8.13282e-10, 2.73256e-8, -2.0499e-7, 5.05345e-7, -2.49295e-7, -3.69
```

```
Δx = 1e-6V[:, 8]
```

In the relative sense, the perturbation Δx is small in the 2-norm.

```
9.9999999985383e-7
```

```
norm((x+Δx)-x)/norm(x)
```

But the image of the perturbed $x + \Delta x$ under H_{15} is much farther away from the image of x .

```
y = Float64[-8.67362e-19, -1.12757e-17, -1.38778e-17, -3.46945e-18, 5.20417e-18,
```

```
· y = A*x
```

```
Δy = Float64[-4.49661e-17, 1.51083e-15, -1.13339e-14, 2.79405e-14, -1.37835e-14,
```

```
· Δy = A*Δx
```

```
1277.699978034397
```

```
· norm(y+Δy-y)/norm(y)
```

It is always true that this relative perturbation is bounded above by the condition number of H_{15} multiplied by the relative perturbation in the domain.

```
true
```

```
· norm(y+Δy-y)/norm(y) ≤ cond(A)*norm((x+Δx)-x)/norm(x)
```

This is because the condition number of the Hilbert matrix is indeed quite large.

```
2.5468136050354794e17
```

```
· cond(A)
```