

Numerical Methods Assignment 4

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1. Polynomial regression and model selection

- 1.1. Assuming $m > n$, write the system of equations in matrix form, identify A , x and b . How many solutions to $Ax = b$ exist in general?

Solution

$$A = \begin{bmatrix} 1 & x_1 & \dots & x_1^m \\ 1 & x_2 & \dots & x_2^m \\ \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & \dots & x_n^m \end{bmatrix}$$

$$x = \begin{bmatrix} a_0 \\ a_1 \\ a_2 \\ \vdots \\ a_m \end{bmatrix}$$

$$b = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$$

...

- 1.2. `fitpoly.m` code:

Solution

```
function [y,coefs,c] = fitpoly(x,xj,yj,n)
```

```
%function fitpoly performs polynomial regression of degree n to the data xj and yj
```

```
%INPUTS
```

```
%-----
```

```
%x - vector to calculate y values for the polynomial generated
```

```
%xj - column vector of known x values
```

```

%yj - column vector of known y values corresponding to the xj vector
%n - degree of the polynomial to generate
%OUTPUTS
%-----
%y - column vector of the values corresponding to the inputted x vector
%coefs - the coefficients corresponding to the solutions to the LSE
%c - the condition number of the regression matrix
%
%-----
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```

```

i = 0:n;
A = xj.^i;
coefs = A\yj;
c = cond(A);
y = x.^i *coefs ;
end

```

...

1.3. Choosing $n = m - 1$ gives $\|e\| = 0$ Why?

Solution

This gives m equations with $m-1$ variables - which can be solved directly. This means the set will pass through all points - giving 0 error. A polynomial of degree k requires $k + 1$ points to generate.

E.g. a linear equation requires 2 points, a quadratic needs 3 points, etc. etc.. ...

1.4. `testfitpoly.m` code:

Solution

```

data = importdata('facebook.dat');
xj = data(:,1);
yj = data(:,2);
maxn=length(xj); %458 as that is n=m
AIC = inf; %initialise to a maximum possible value
x = xj;
for n=1:maxn
    [y,coefs,c] = fitpoly(x,xj,yj,n);
    ej = y - yj;
    AICn = 2*n + 458*log(sum(ej.^2));
end

```

```

        if(AICn > AIC)
            AIC = AICn;
            break;
        end
        AIC = AICn;

    end
    hold on
    plot(x,y,'r-')
    plot(xj,yj,'b.')
    title("Polynomial regression on Facebook search data")
    xlabel("Number of years since 1 January 2007")
    ylabel("Normalised weekly search volume counts for Facebook")
    legend(["Polynomial corresponding to minimum AIC, n = " + num2str(n)],"Given Fac
    hold off
    %%1.6
    figure
    x = linspace(min(xj),11,500)';
    [y,~,~] = fitpoly(x,xj,yj,n);
    plot(x,y,'b',x(end),y(end),'r.')
    title("Prediction for 2017")
    xlabel("Number of years since 1 January 2007")
    ylabel("Normalised weekly search volume counts for Facebook")

    %%1.7
    [y,coefs,c] = fitpoly(xj,xj,yj,20);

```

...

- 1.5. Plot the data and n th degree polynomial over the same time period. Label x axis as "number of years since 1 January 2007"

Solution

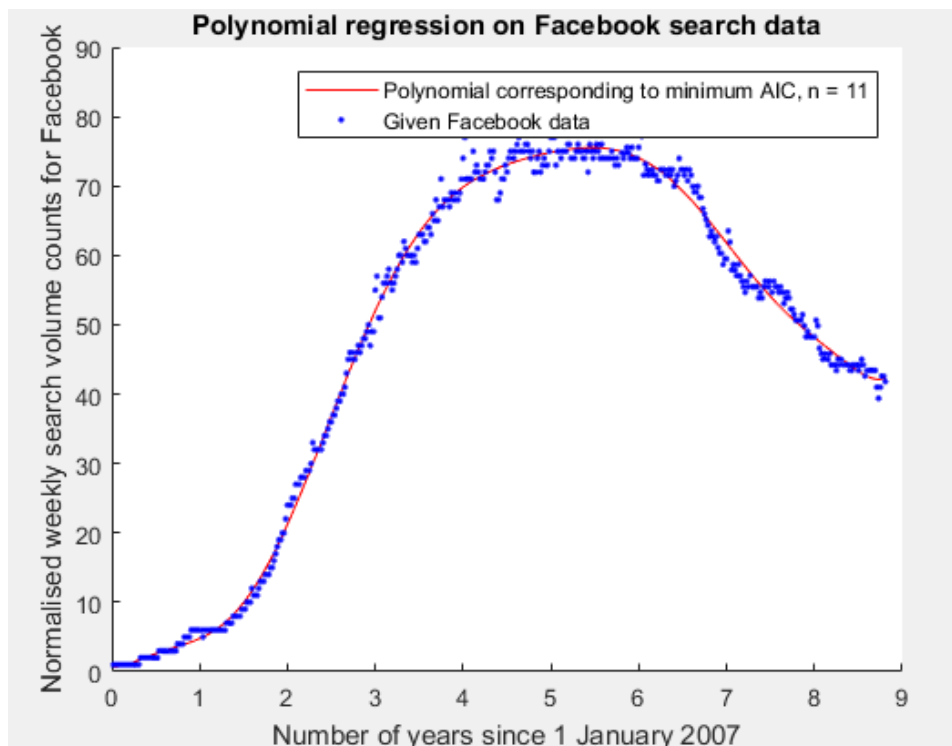
Above is the n -th degree polynomial over the same period ...

- 1.6. Could the model reasonably predict Facebook search volume today in September 2017? Why or why not?

Solution

Firstly the graph generated by the model for 2017 is below:

From this it appears it is not very effective in predicting present or future search volume. This would have several reasons including: A lot of people use the Facebook app, many people save bookmarks rather than using google to find it, and facebook's use may have even peaked. ...



- 1.7. Fit a degree 20 polynomial to the dataset. With reference to the outputs of `fitpoly`, explain why Matlab throws a warning message here.

Solution

The error message states:

> In `fitpoly` (line 23)

In `testfitpoly` (line 36)

Warning: Rank deficient, **rank** = 7, tol = 2.743044e+06.

This error corresponds to the line `coefs = A\y`, where `A` is the matrix `A` shown in 1.1. The issue is `A` is rank deficient. It has rank 7, when it needs rank 21 (to give solutions). The issue arises as a number of the values in the rows of `A` become too small for machine precision and become 0 throughout. ...

2. Solution of the heat equation using Jacobi iteration

- 2.1. Referring to the relevant theorem - explain why Jacobi iteration is guaranteed to converge to the solution $Au = b$ when $\alpha > 0$

Solution

Theorem 5.32 states that Jacobi iteration will converge if the matrix `A` is diagonally dominant. If $\alpha > 0$ then the terms on the

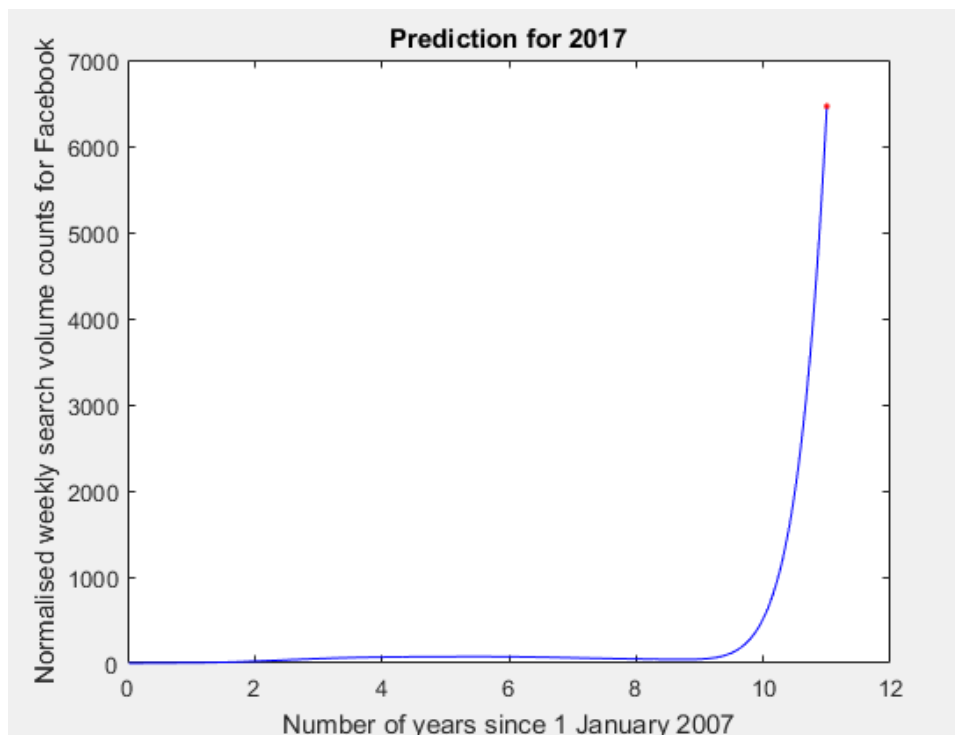


Figure 1: Trend from extrapolation

diagonal will have absolute value $2 + \epsilon$ for $\epsilon = \alpha h^2$. The matrix is diagonally dominant if

$$|-2 - \alpha h^2| \geq |1| + |1|$$

So for $\alpha > 0$

$$2 + \epsilon > 2$$

So the matrix is diagonally dominant.

...

- 2.2. Using the information given, write down a formula for the elements of the Jacobi iterate $u^{(k)}$ in the form

$$u_i^{(k)} = u_i^{(k-1)} - ?, \quad i = 2, \dots, N-1$$

Solution

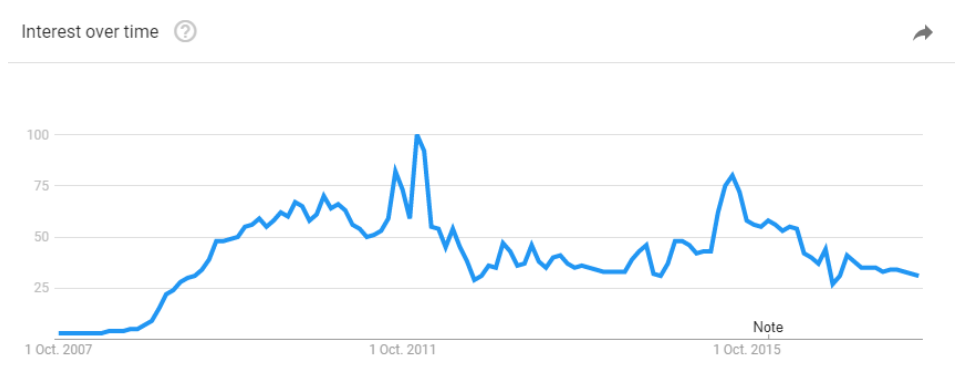


Figure 2: True trend from Google trends

Given the normal form: $u^{(k)} = u^{(k-1)} - D^{-1}r^{(k-1)}$

$$\begin{aligned} u^{(k)} &= u^{(k-1)} - D^{-1}r^{(k-1)} \\ &= u_i^{(k-1)} - \frac{-1}{2 + \alpha h^2} r_i^{(k-1)} \\ u_i^{(k)} &= u_i^{(k-1)} + \frac{r_i^{(k-1)}}{2 + \alpha h^2} \end{aligned}$$

...

2.3. Write a function `solveheat.m` to solve $Au = b$ using Jacobi iteration.

Solution

As below:

```
function [u,x,rnorm] = solveheat(alpha,func,L,u0,uL,N,tol,maxits)
%Function solveheat
%solves the heat PDE in one dimension in the form  $d^2u/dx^2 - \alpha * u = -f(x)$ 
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%INPUTS
%-----
%alpha - the value corresponding to alpha in the PDE
%func - the function  $f(x)$  to equate to
%L - length of the 'heated rod'
%u0 - initial condition corresponding to the LHS of the rod
%uL - initial condition corresponding to the RHS of the rod
%N - number of points to solve over
%tol - solution tolerance - if the equation varies by less than this over
```

```

%          an iteration , break.
%OUTPUTS
%-----
%u      - the solution to the PDE at x
%x      - the vector of N evenly distributed points between 0 and L
%rnorm - the value

```

```

h = L/(N-1);
u=zeros(N,1);
u(1) = u0;
u(N) = uL;
j=1:N;
i=2:N-1;
x = (j-1)*h;
fx = func(x);
r = zeros(N,1);
Dinv = 1/(2+alpha*h^2);
for iteration=1:maxits
    r(i)= u(i-1) - (2+alpha*h^2)*u(i) + u(i+1) + h^2 * fx(i);
    u(i) = u(i) + Dinv * r(i);
    rnorm = norm(r);
    if(rnorm<tol)
        disp(iteration);
        break;
    end
end
end

end

```

...

2.4. Considering the BVP given, verify the exact solution is

$$u(x) = \frac{1 + \frac{1}{2} \sin(1)}{\sinh(1)} \sinh(x) - \frac{1}{2} \sin(x)$$

Solution

$$\begin{aligned}
\frac{\partial^2 u}{\partial x^2} - u &= \sin(x) \\
\lambda^2 - \lambda &= 0 \\
\lambda &= 0, \quad \lambda = 1 \\
u &= c_1 e^x + c_2 e^{-x} + d \sin(x) \\
\frac{\partial^2 u}{\partial x^2} &= c_1 e^x + c_2 e^{-x} - d \sin(x) \\
\frac{\partial^2 u}{\partial x^2} - u &= -2d \sin(x) = \sin(x) \implies d = \frac{-1}{2} \\
u &= c_1 e^x + c_2 e^{-x} - \frac{\sin(x)}{2}
\end{aligned}$$

Applying BCs:

$$\begin{aligned}
u &= c_1 e^x + c_2 e^{-x} - \frac{\sin(x)}{2} \\
u(0) = 0 &= c_1 e^0 + c_2 e^0 - \frac{\sin(0)}{2} \\
&= c_1 + c_2 \implies c_2 = -c_1 \\
u(1) = 1 &= c_1 e^1 - c_1 e^{-1} - \frac{\sin(1)}{2} \\
c_1(e^1 - e^{-1}) &= 1 + \frac{\sin(1)}{2} \\
\implies c_1 &= \frac{1 + \frac{1}{2} \sin(1)}{(e^1 - e^{-1})}
\end{aligned}$$

Which gives

$$\begin{aligned}
u &= -\frac{1 + \frac{1}{2} \sin(1)}{(e^1 - e^{-1})} e^x + \frac{1 + \frac{1}{2} \sin(1)}{(e^1 - e^{-1})} e^{-x} - \frac{\sin(x)}{2} \\
&= -\frac{1 + \frac{1}{2} \sin(1)}{(e^1 - e^{-1})} (e^x + e^{-x}) - \frac{\sin(x)}{2} \\
&= -\frac{1 + \frac{1}{2} \sin(1)}{\sinh(1)} \sinh(x) - \frac{1}{2} \sin(x)
\end{aligned}$$

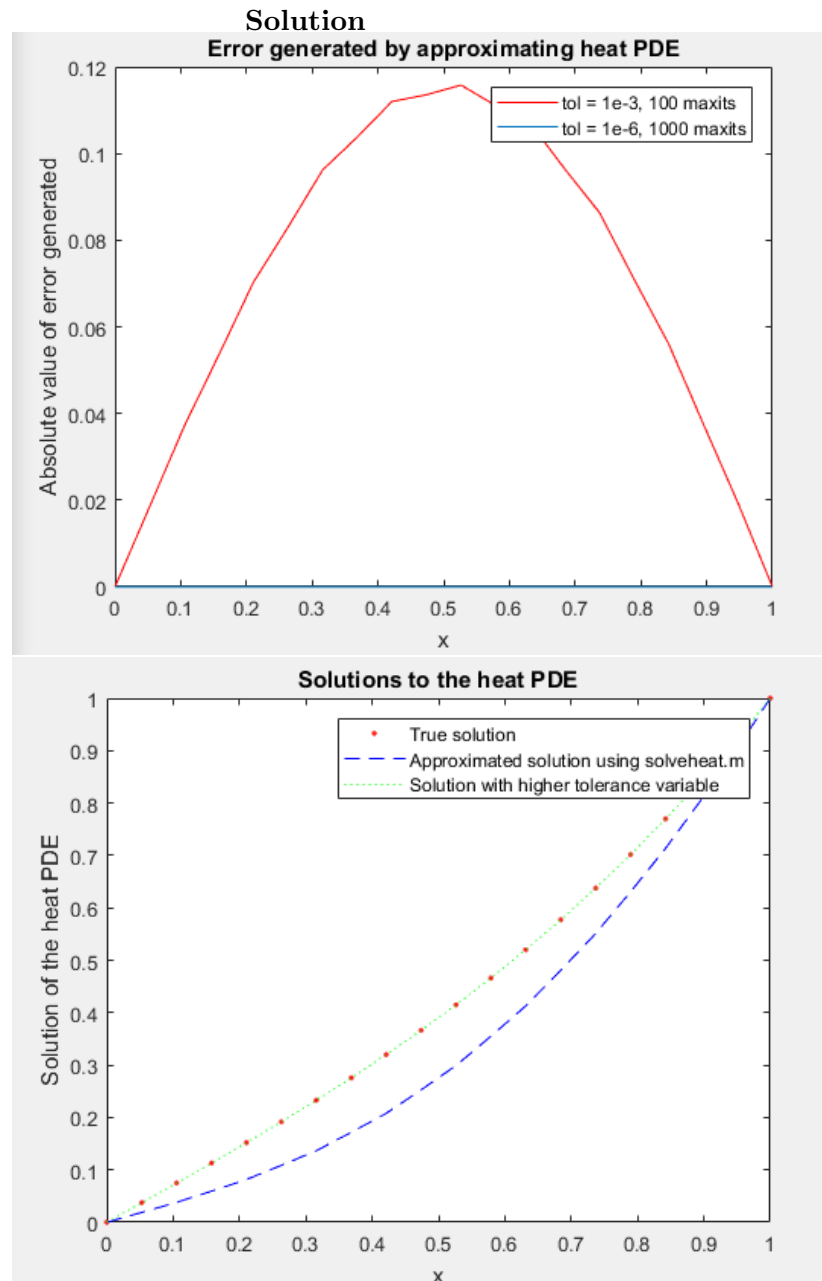
...

2.5. Write `testsolveheat.m` and create two plots:

- a. Showing exact solution and approximate solutions as functions of x

- b. Another showing the absolute value of the difference between those two.

Make sure plots are labelled.



- 2.6. Add code to `testsolveheat.m` that uses Jacobi iteration to solve the heat equation with given ICs and BCs

Solution

The entirety of `testsolveheat.m` is below

```
fun = @(x) (1+0.5*sin(1))/sinh(1) *sinh(x) - 0.5*sin(x);
func = @(x) -sin(x);
```

```
%%Iterate solutions to Heat PDE
```

```
[u,x,rnorm] = solveheat(1,func,1,0,1,20,1e-3,100);
```

```
[u2,~,~] = solveheat(1,func,1,0,1,20,1e-6,1000);
```

```
plot(x,fun(x),'r.',x,u,'b--',x,u2,'g:')
```

```
title("Solutions to the heat PDE")
```

```
legend("True solution","Approximated solution using solveheat.m","Solution with higher resolution")
```

```
xlabel("x");
```

```
ylabel("Solution of the heat PDE");
```

```
%%Error generated
```

```
figure
```

```
diff = abs(fun(x)-u);
```

```
diff2 = abs(fun(x)-u2);
```

```
plot(x,diff,'r',x,diff2,'b');
```

```
title("Error generated by approximating heat PDE")
```

```
legend("tol = 1e-3, 100 maxits","tol = 1e-6, 1000 maxits")
```

```
xlabel("x");
```

```
ylabel("Absolute value of error generated");
```

```
%2.6
```

```
N=100;
```

```
L=1;
```

```
alpha = 1;
```

```
u0 = 0;
```

```
uL = 1;
```

```
f = @(x) 80*exp(-(x-0.4).^2/0.001);
```

```
tol = 1e-3;
```

```
maxits = 10000;
```

```
[u3,x,rnorm] = solveheat(alpha,f,L,u0,uL,N,tol,maxits);
```

```
figure
```

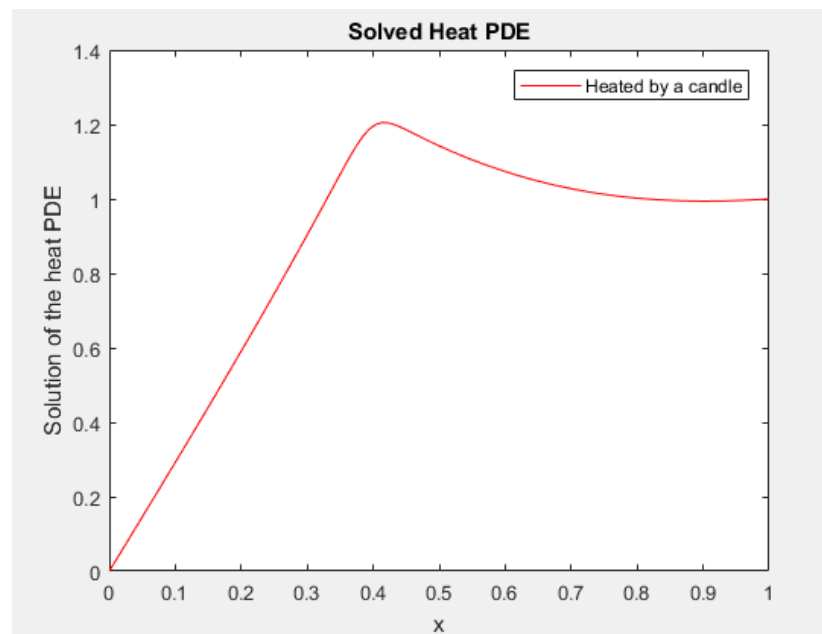
```
plot(x,u3,'r-');
```

```
title("Solved Heat PDE")
```

```
legend("Heated by a candle")
```

```
xlabel("x");
```

```
ylabel("Solution of the heat PDE");
```



...