

In [1]:

*#Problem 1*

**using** PyPlot, JuMP, Gurobi

$X = 4 + \text{randn}(2,50)$  *# generate 50 random points*

$t = \text{linspace}(0,2\pi,100)$  *# parameter that traverses the circle*

*# model solves for radius squared due to needing PSD*

*# and center point*

$m = \text{Model}(\text{solver}=\text{GurobiSolver}(\text{OutputFlag}=0))$

**@variable**( $m$ ,  $r^2 \geq 0$ )

**@variable**( $m$ ,  $x[1:2]$ )

**@constraint**( $m$ ,  $\text{pos}[i = 1:50]$ ,  $\text{dot}((X[:, i] - x), (X[:, i] - x)) \leq r^2$ )

**@objective**( $m$ ,  $\text{Min}, r^2$ )

$\text{solve}(m)$

*# model solves for radius squared due to needing PSD*

$r = \sqrt{\text{getobjectivevalue}(m)}$

$x_1 = \text{getvalue}(x[1])$

$x_2 = \text{getvalue}(x[2])$

$\text{println}(\text{"Radius is: "}, r)$

$\text{println}(\text{"Center point coordinates are: ("}, x_1, \text{"}, x_2, \text{"})")$

*# plot circle radius r with center (x1,x2)*

$\text{plot}(x_1 + r \cdot \cos(t), x_2 + r \cdot \sin(t))$

$\text{scatter}(X[1,:], X[2:], \text{color}=\text{"black"})$  *# plot the 50 points*

$\text{axis}(\text{"equal"})$  *# make x and y scales equal*

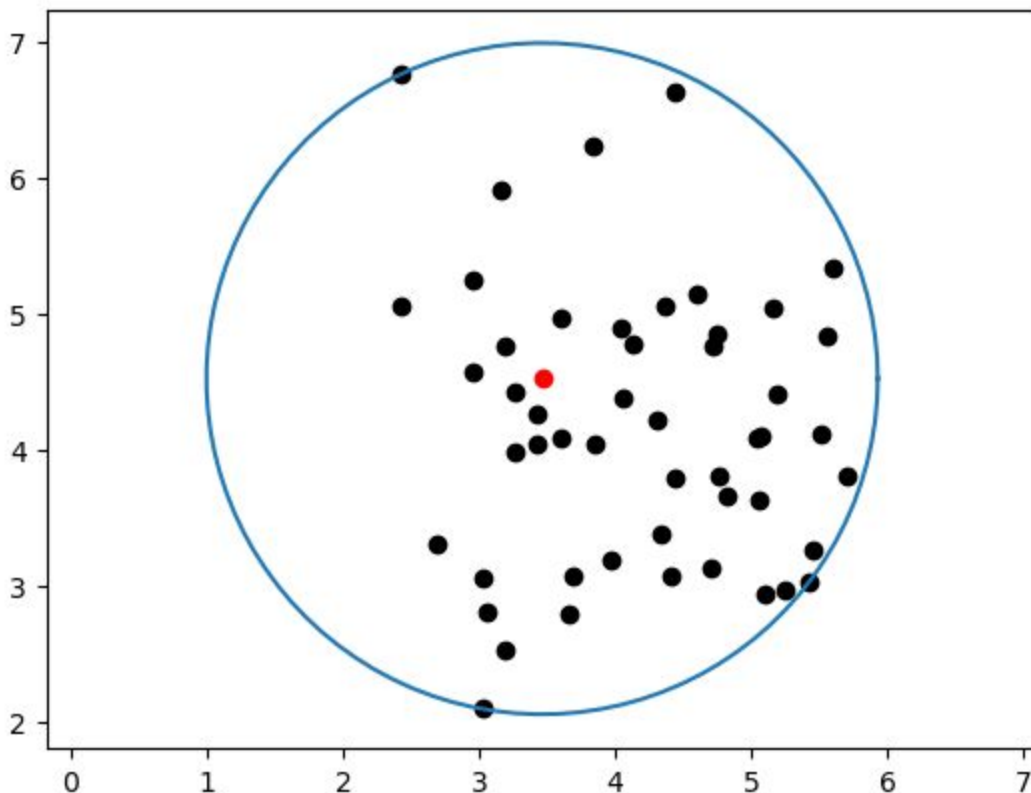
*# plot the center*

$\text{scatter}(x_1, x_2, \text{color}=\text{"red"})$ ;

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Radius is: 2.465731708804874

Center point coordinates are: (3.462036947827601,4.522762785474279)



*#Problem 2a*

$Q = \begin{bmatrix} 2 & 4 & -3 \\ 4 & 2 & -3 \\ -3 & -3 & 9 \end{bmatrix}$

*#v = [x y z]*

3×3 Array{Int64,2}:

2 4 -3

4 2 -3

-3 -3 9

*#Problem 2b*

$(L, U) = \text{eig}(Q)$

*#Here we see that one of the eigenvalues of Q is -2, so Q is not positive definite, meaning it is not an ellipsoid.*

Out[147]:

$([-2.0, 3.0, 12.0], [0.707107 \ -0.57735 \ -0.408248; -0.707107 \ -0.57735 \ -0.408248; 0.0 \ -0.57735 \ 0.816497])$

In [148]:

*#Problem 2c*

*#Q equals  $U \cdot \text{diagm}(L) \cdot U'$  so  $\text{diagm}(L)$  can be written as the difference of the matrices*

$M = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 12 \end{bmatrix}$

$N = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

$\text{println}(\text{"U*M*U' - U*N*U'": }, U \cdot M \cdot U' - U \cdot N \cdot U')$

$\text{println}(\text{"U: "}, U)$

$U \cdot M \cdot U' - U \cdot N \cdot U': [2.0 \ 4.0 \ -3.0; 4.0 \ 2.0 \ -3.0; -3.0 \ -3.0 \ 9.0]$

$U: [0.707107 \ -0.57735 \ -0.408248; -0.707107 \ -0.57735 \ -0.408248; 0.0 \ -0.57735 \ 0.816497]$

$\text{println}(\text{"U*U'": }, U \cdot U')$

$U \cdot U': [1.0 \ -8.32667\text{e-}17 \ 1.11022\text{e-}16; -8.32667\text{e-}17 \ 1.0 \ 1.11022\text{e-}16; 1.11022\text{e-}16 \ 1.11022\text{e-}16 \ 1.0]$

Handwritten mathematical expressions:

$$A = U \begin{bmatrix} 0 & 0 & 0 \\ 0 & \sqrt{3} & 0 \\ 0 & 0 & \sqrt{12} \end{bmatrix} U^T$$
$$B = U \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} U^T$$
$$v = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$A = U \cdot \begin{bmatrix} 0 & 0 & 0 \\ 0 & \sqrt{3} & 0 \\ 0 & 0 & \sqrt{12} \end{bmatrix} \cdot U'$

$B = U \cdot \begin{bmatrix} \sqrt{2} & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \cdot U'$

$\text{println}(\text{"A'*A + B'*B": }, A' \cdot A + B' \cdot B)$

$A' \cdot A + B' \cdot B: [4.0 \ 2.0 \ -3.0; 2.0 \ 4.0 \ -3.0; -3.0 \ -3.0 \ 9.0]$

$$v = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

Now we have:

$$\begin{aligned} \|Av\|^2 - \|Bv\|^2 &= (Av)^T(Av) - (Bv)^T(Bv) \\ &= v^T A^T A v - v^T B^T B v \\ &= v^T (A^T A - B^T B) v \\ &= v^T \begin{bmatrix} 0 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 12 \end{bmatrix} v - v^T \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} v \\ &= v^T Q v \end{aligned}$$

In [ ]:

#Problem 2d

Constraint (1) can be written as:  $v^T Q v \leq 1$

Since  $Q = U \begin{bmatrix} -2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 12 \end{bmatrix} U^T$ , let  $z = U^T v$

$U$  is orthogonal so:

$$z = U^T v = U^{-1} v$$

$$v = U z$$

$$x^2 + y^2 + z^2 = \|v\|^2 = v^T v = (U z)^T U z = z^T U^T U z = z^T z = \|z\|^2$$

the constraint, then, can be written as  $z^T \begin{bmatrix} -2 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 12 \end{bmatrix} z \leq 1$

Since  $\|v\|^2 = \|z\|^2$ , assume we want  $\|v\|^2 = k$  where  $k$

is an arbitrary large number. Then the problem can be converted to:

$$z_1^2 + z_2^2 + z_3^2 = k \quad (1)$$

$$-2z_1^2 + 3z_2^2 + 12z_3^2 \leq 1 \quad (2)$$

To solve, set one element as fixed and solve for the other two. Then use  $v = U z$  to find  $(x, y, z)$  that satisfies (1) with arbitrarily large  $k$ .

In [136]:

```
#problem 3
```

```
using PyPlot, JuMP, Gurobi
```

```
raw = readcsv("lasso_data.csv");
```

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

```
y = raw[:,2]
```

```
figure(figsize=(8,4))
```

```
p = plot(x, y, ".")
```

```
axis([.1,.6,-1.5,1.5])
```

```
grid()
```

```
function f(k)
```

```
k = k
```

```

n = length(x)
A = zeros(n,k+1)
for i in 1:n
    for j = 1:k+1
        A[i,j] = x[i]^(k+1-j)
    end
end

```

```

m = Model(solver = GurobiSolver(OutputFlag=0))

```

```

@variable(m, u[1:k+1])
@objective(m, Min, sum((y-A*u).^2))

```

```

status = solve(m)
uopt = getvalue(u)
println(status)
println(getobjectivevalue(m))
println(getvalue(u))

```

```

npts = 10000
xfine = linspace(0,10,npts)
ffine = ones(npts)

```

```

for j = 1:k
    ffine = [ffine .*xfine ones(npts)]
end

```

```

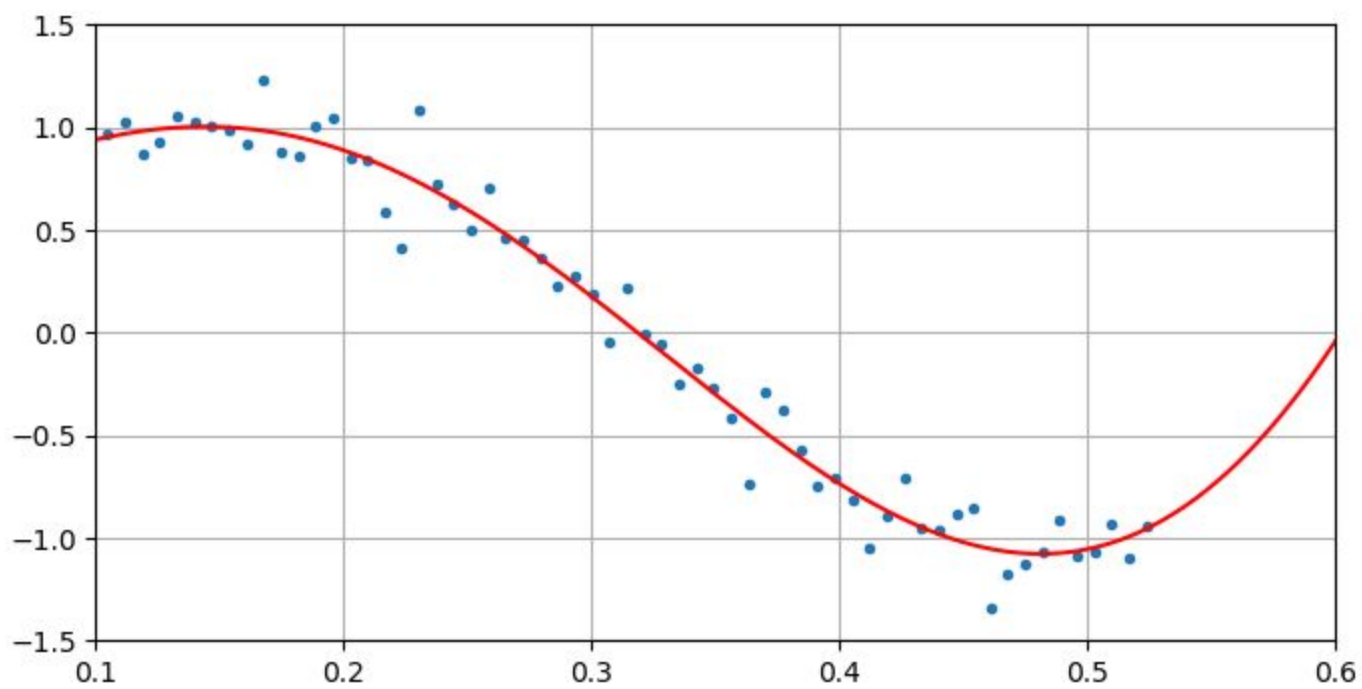
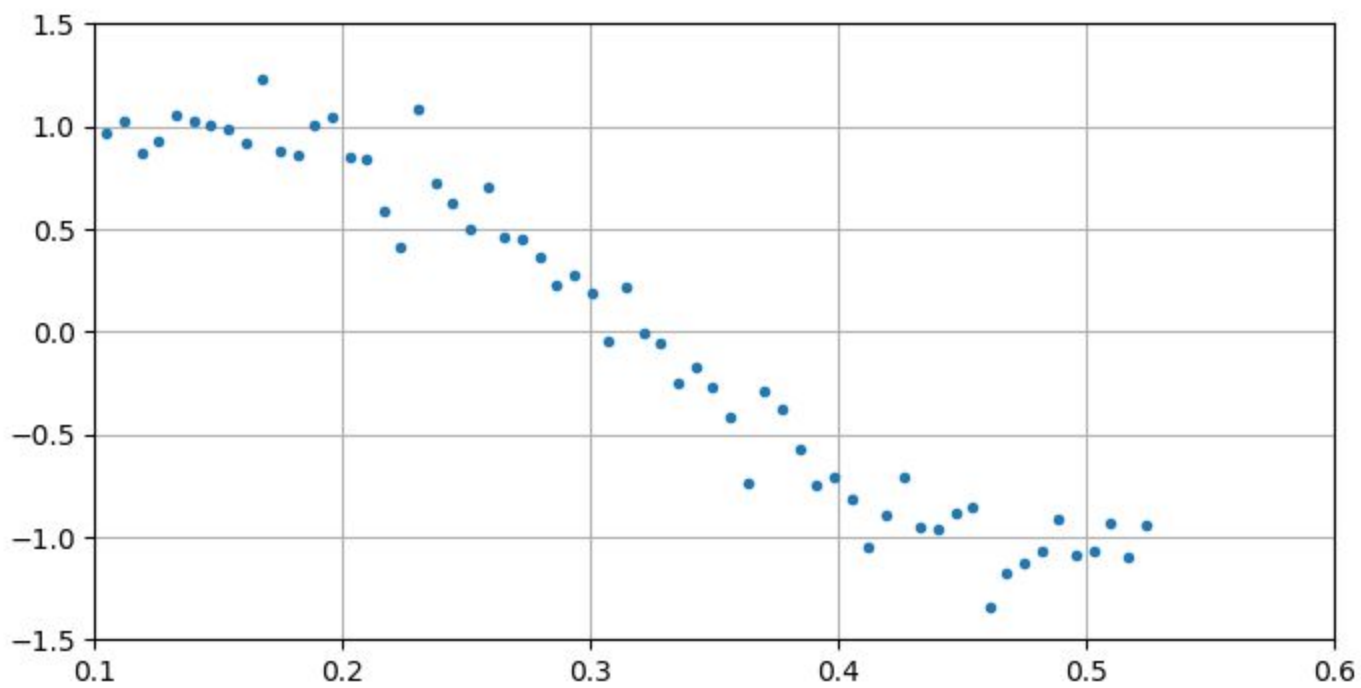
yfine = ffine * uopt
figure(figsize=(8,4))
plot(x,y, ".")
plot(xfine,yfine, "r-")
axis([.1,.6,-1.5,1.5])
grid()

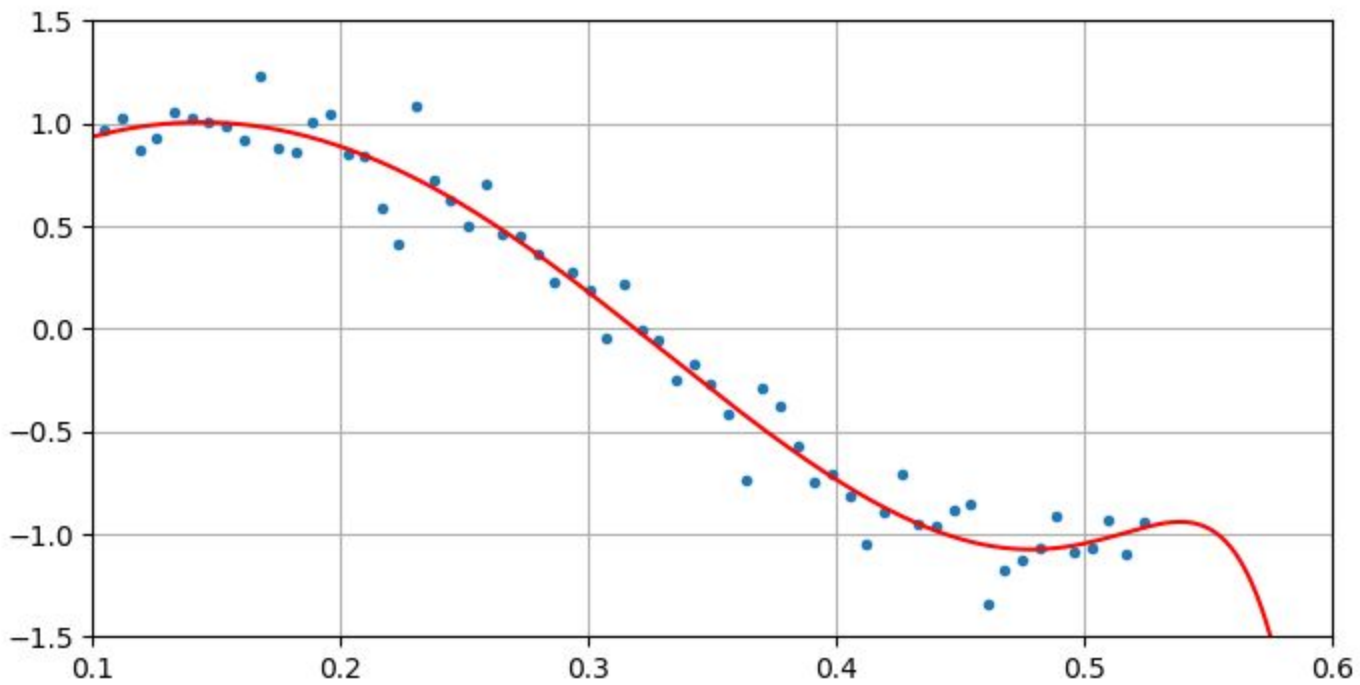
```

```

end
f(5)
f(15)

```





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Optimal

1.0144487799939412

[-320.872, 619.059, -333.419, 41.1237, 2.18777, 0.584456]

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Optimal

1.0135488280265221

[-3.37904e5, 1.86649e5, 74678.9, -4481.85, -17029.3, -8090.89, -745.502, 1319.48, 774.64, 312.963, -144.165, -31.6809, 18.4594, -40.3935, 10.9502, 0.228989]

In [ ]:

*# Here we see that the magnitudes of the coefficients are very large.*

In [101]:

*#Problem 3b*

k = 15

n = length(x)

A = zeros(n, k+1)

**for** i = 1:n

**for** j = 1:k+1

        A[i,j] = x[i]^(k+1-j)

**end**

**end**

**using** JuMP, Gurobi

$\lambda = 1/(10^6)$

```
m = Model(solver=GurobiSolver(OutputFlag=0))
```

```
@variable(m, u[1:k+1])
```

```
@objective(m, Min, sum((y-A*u).^2) +  $\lambda$ *sum(u.^2))
```

```
status = solve(m)
```

```
uopt = getvalue(u)
```

```
println(status)
```

```
println("uopt:", uopt)
```

```
println("Error:", sum((y-A*uopt).^2))
```

**using** PyPlot

```
npts = 100
```

```
xfine = linspace(0, 0.6, npts)
```

```
ffine = ones(npts)
```

```
for j = 1:k
```

```
    ffine = [ffine.*xfine ones(npts)]
```

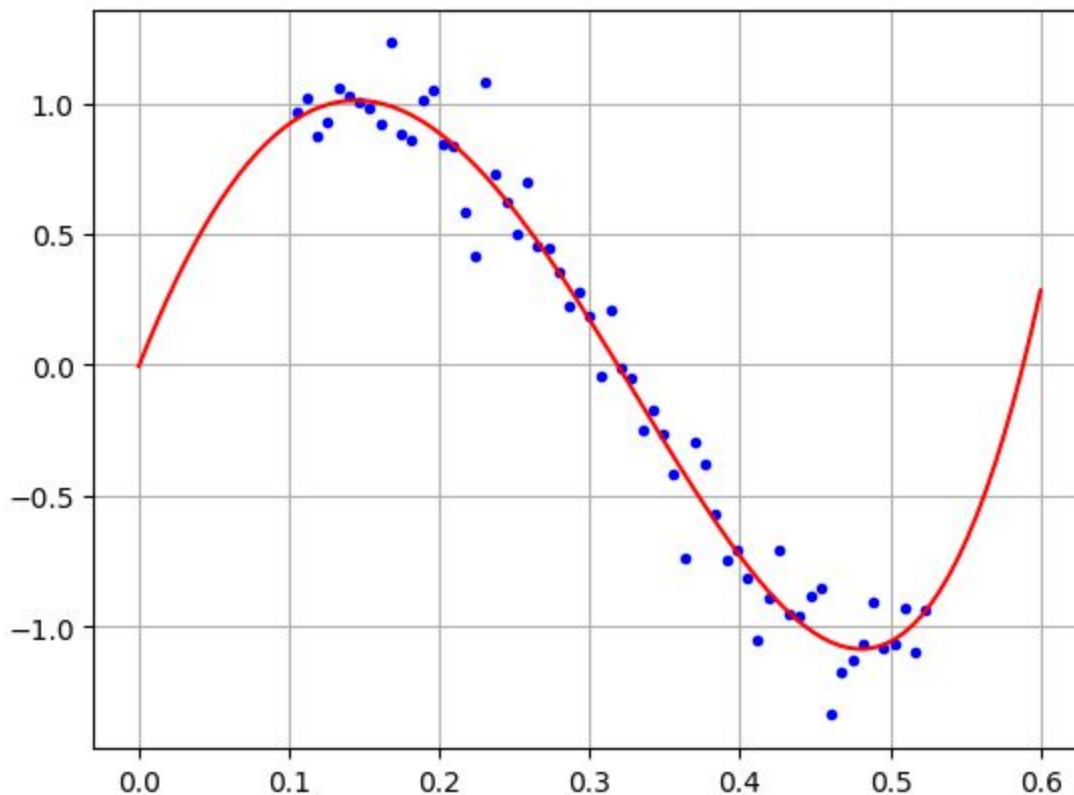
```
end
```

```
yfine = ffine * uopt
```

```
plot(x,y,"b.")
```

```
plot(xfine,yfine,"r-")
```

```
grid()
```





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Optimal

```
uopt:[-0.59465, -0.975681, -1.54994, -2.35464, -3.35103, -4.29544, -4.51095, -2.58953, 3.71018, 16.501, 33.9216, 42.0711, 11.6188, -55.0281, 14.6015, -0.00540095]
```

Error:1.0168419826436905

*# Here we see that using L2 regularization causes the error to get larger while the magnitudes of the coefficients get smaller*

---

In [119]:

*#Problem 3c*

```
input = readcsv("lasso_data.csv")
```

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

```
y = input[:, 2]
```

```
k = 15
```

```
n = length(x)
```

```
A = zeros(n, k+1)
```

```
for i = 1:n
```

```
    for j = 1:k+1
```

```
        A[i,j] = x[i]^(k+1-j)
```

```
    end
```

```
end
```

```
λ_array = []
```

```
error_array = []
```

```
nonzero_items = []
```

```
using JuMP, Gurobi
```

```
for λ in [1, 0.1, 0.01, 0.001, 0.0001, 0.00001, 0.000001, 0.0000001, 0.00000001]
```

```
    m = Model(solver=GurobiSolver(OutputFlag=0))
```

```
    # l1-norm
```

```
    @variable(m, t[1:k+1])
```

```
    @variable(m, u[1:k+1])
```

```
    @constraint(m, u .<= t)
```

```
    @constraint(m, -t .<= u)
```

```
    @objective(m, Min, sum((y-A*u).^2) + λ*sum(t))
```

```
    status = solve(m)
```

```

uopt = getvalue(u)

push!(λ_array, log10(λ))
push!(error_array, sum((y-A*(getvalue(u))).^2))

nz_items = 0
# print non-zero items
for i = 1:k+1
    if abs(uopt[i]) >= 1/(10^5)
        # println(i, " : ", uopt[i])
        nz_items = nz_items + 1
    end
end

# println("nonzero items:", nz_items)
push!(nonzero_items, nz_items)

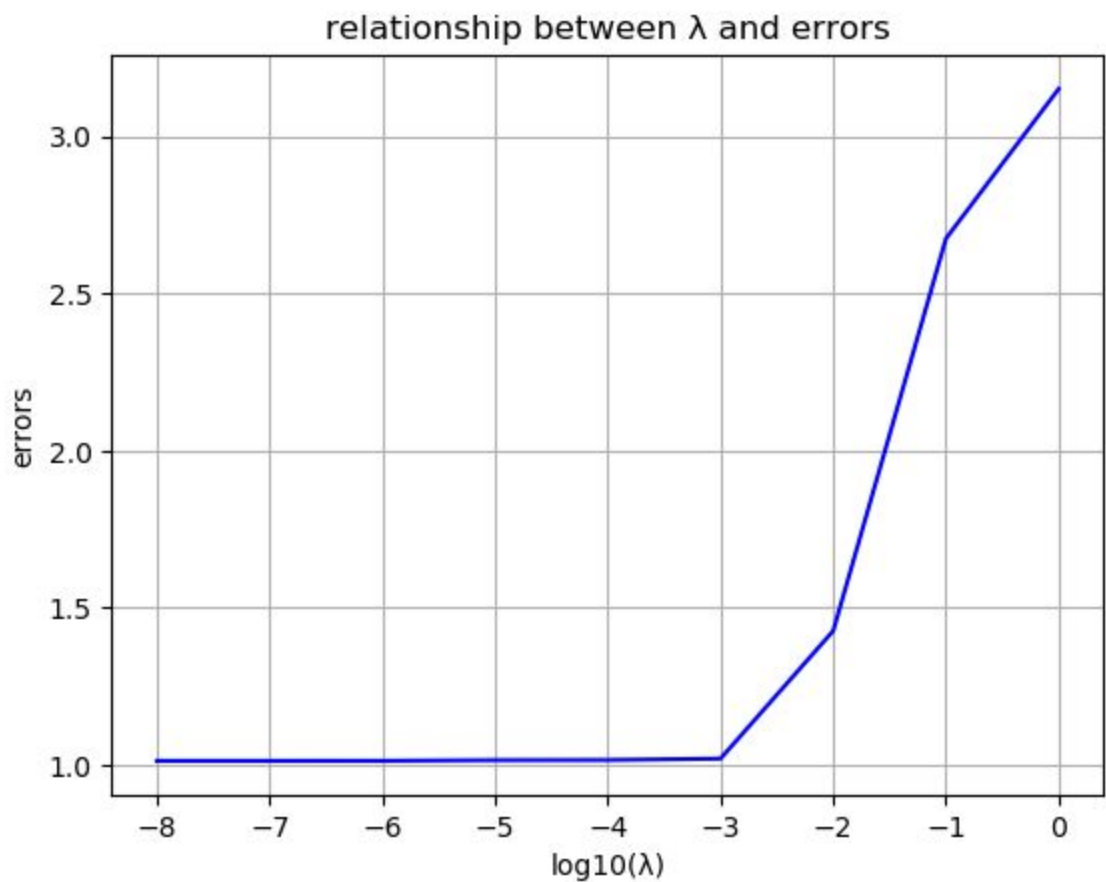
```

end

```

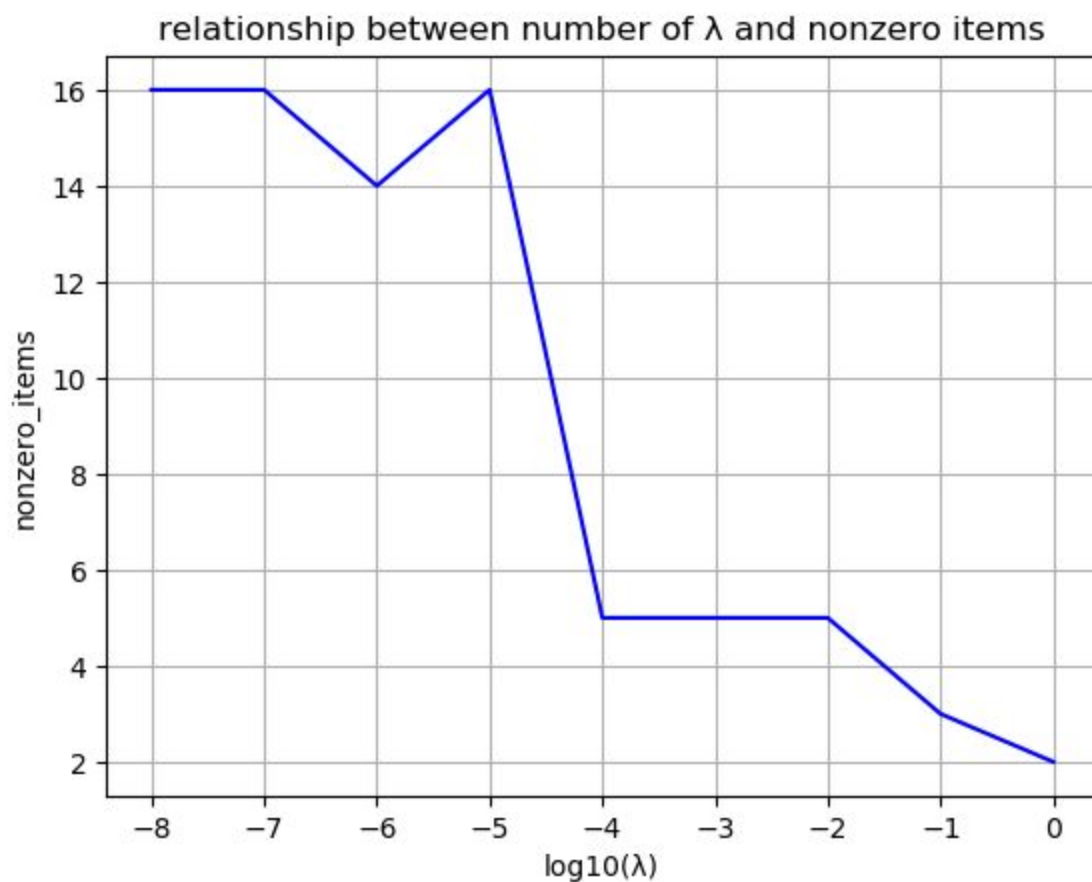
using PyPlot
Plots.default(overwrite_figure=false)
title("λ and error corelation")
plot(λ_array, error_array, "b-")
xlabel("log10(λ)")
ylabel("errors")
grid()

```



In [118]:

```
title("number of  $\lambda$  and nonzero items correlation")
plot( $\lambda\_array$ , nonzero_items, "b-")
xlabel("log10( $\lambda$ )")
ylabel("nonzero_items")
grid("on")
```



In [125]:

```
# From the two graphs above, we see that when  $\lambda = 10^{-3}$  we have five nonzero
# items and the error is 1. When  $\lambda$  gets smaller, the error will not change but
# there will be more nonzero items. When  $\lambda$  gets bigger, the error will increase
# but there will be less nonzero items.  $10^{-3}$  is a good middle ground. I use this
# value of  $\lambda$  in the following.
```

In [126]:

```

input = readcsv("lasso_data.csv")
x = input[:, 1]
y = input[:, 2]

k = 15

n = length(x)
A = zeros(n, k+1)
for i = 1:n
    for j = 1:k+1
        A[i,j] = x[i]^(k+1-j)
    end
end

using JuMP, Gurobi

λ = 0.001

m = Model(solver = GurobiSolver(OutputFlag=0))

# l1-norm
@variable(m, t[1:k+1])
@variable(m, u[1:k+1])
@constraint(m, u .<= t)
@constraint(m, -t .<= u)
@objective(m, Min, sum((y-A*u).^2) + λ*sum(t))

status = solve(m)

uopt = getvalue(u)

println(status)
println("uopt: ", uopt)
println("error: ", sum((y-A*(getvalue(u))).^2))

nz_items = 0
# print non-zero items
for i = 1:k+1
    if abs(uopt[i]) >= 1/(10^5)
        println(i, " : ", uopt[i])
        nz_items = nz_items + 1
    else
        println(i, " : ", 0)
    end
end

println("nonzero items:", nz_items)

```

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Optimal

uopt: [1.50878e-9, 2.97466e-9, 5.82115e-9, 1.12857e-8, 2.16288e-8, 4.09081e-8, 7.65213e-8, 1.43863e-7, 2.88423e-7, 7.57818e-7, 41.1878, 50.3653, 1.51781e-7, -48.5432, 13.0343, 0.119144]

error: 1.020887943014919

1 : 0

2 : 0

3 : 0

4 : 0

5 : 0

6 : 0

7 : 0

8 : 0

9 : 0

10 : 0

11 : 41.18778432016339

12 : 50.36533671321371

13 : 0

14 : -48.54323558552761

15 : 13.034344853120968

16 : 0.11914408814181378

nonzero items:5