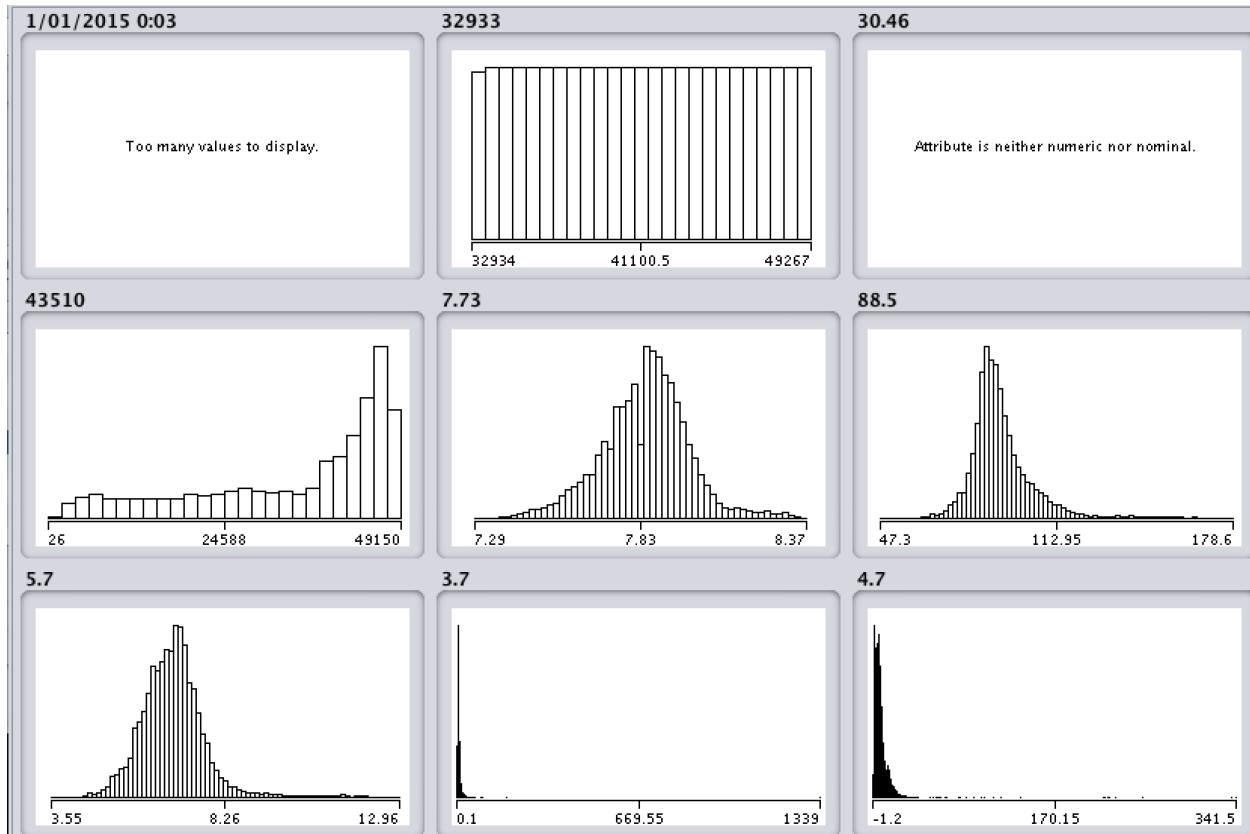


P1Q2.

This csv file contains nine set of variables, which is recorded from 0:03 of 1<sup>st</sup> January 2015 until 23:33 of 31<sup>st</sup> Dec 2015. It took data every half hours during all year and data index starts from 32933 and ends to 49267.



From the histogram graph, the fourth set demonstrated a right-skewed distribution from 26 to 49150 with mean 34670. The fifth, sixth and seventh set of data all demonstrate as a normal distribution, whose mean and standard deviation are 7.846 with 0.142, 92.423 with 12.206, 6.838 with 0.89 respectively. Whereas the last two graph are left-skewed distribution, which concentrate on minimum value 0.1 and -1.2 respectively.

From the visualization part, we are trying to find the relationship between those data set.

- The dataset 1 (Timeline) and dataset 2 (Index) has a clearly positive linear distribution.
- The dataset 3 (30.46) and dataset 1 (Time)/dataset 2 (Index) are demonstrate a normal distribution. It initially increasing with time increasing, all the way until mid-year reach the maximum number, and then decreasing to the end of the year.
- The dataset 6 (88.5) and dataset 7 (5.7) are demonstrated to be a positive correlation. As visualizing shows, increasing number of dataset6 will results to an increasing number of dataset 7.

P1Q6.

```
function out = reverse(x, y)
```

```
    out = zeros(numel(x),1);
```

```
    if mod(numel(x), y) == 0
```

```
        loop_num = numel(x)/y ;
```

```
        loop_count = 0;
```

```
        while(loop_num > 0)
```

```
            for i = 1 : y
```

```
                out(i + loop_count*y) = x(numel(x) + i - (loop_count+1)*y);
```

```
            end
```

```
            loop_count = loop_count+1;
```

```
            loop_num = loop_num - 1;
```

```
        end
```

```
    else
```

```
        loop_num = uint8(mod(numel(x), y)) - 1;
```

```
        loop_count = 0;
```

```
        while(loop_num > 0)
```

```
            for i = 1 : y
```

```
                out(i + loop_count*y) = x(numel(x) + i - (loop_count+1)*y);
```

```
            end
```

```
            loop_count = loop_count+1;
```

```
            loop_num = loop_num - 1;
```

```
        end
```

```
        for i = 1:mod(numel(x),y)
```

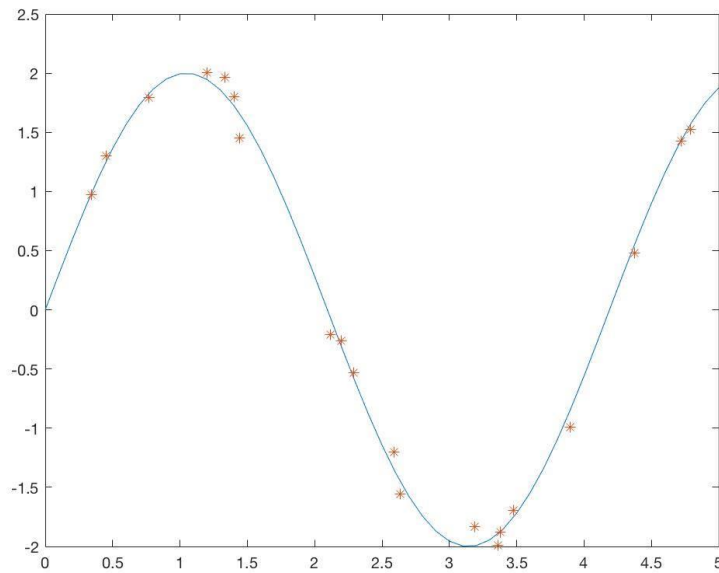
```
            out(((loop_count)*y) + i) = x(i);
```

```
        end
```

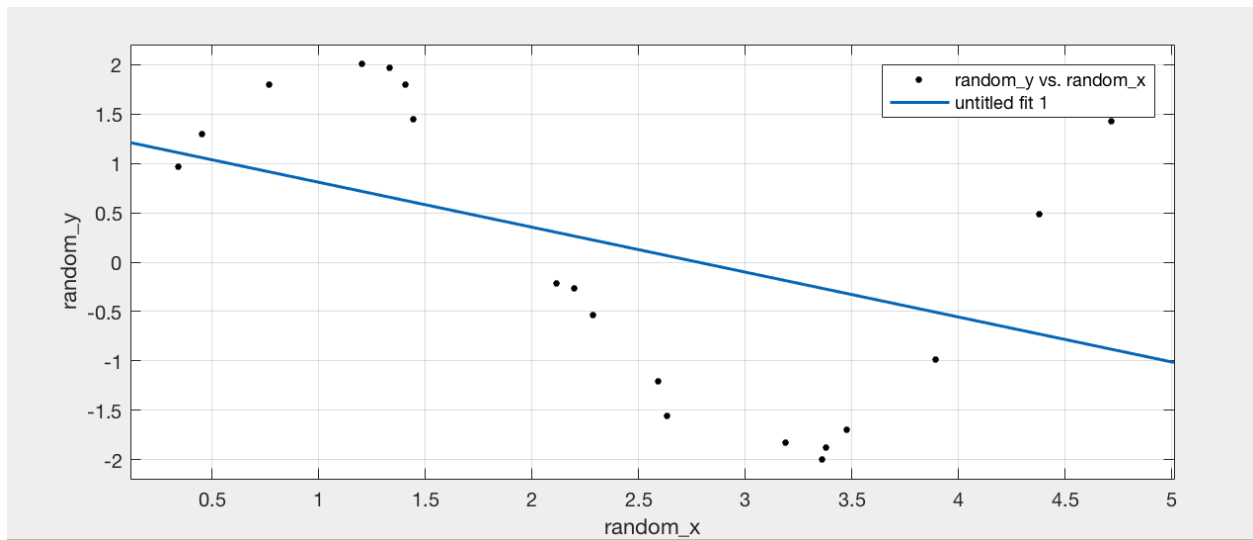
```
    end
```

```
end
```

P2Q1.



(Figure:  $y=2\sin(1.5x)$  and 20 random points with 0.1 variance)



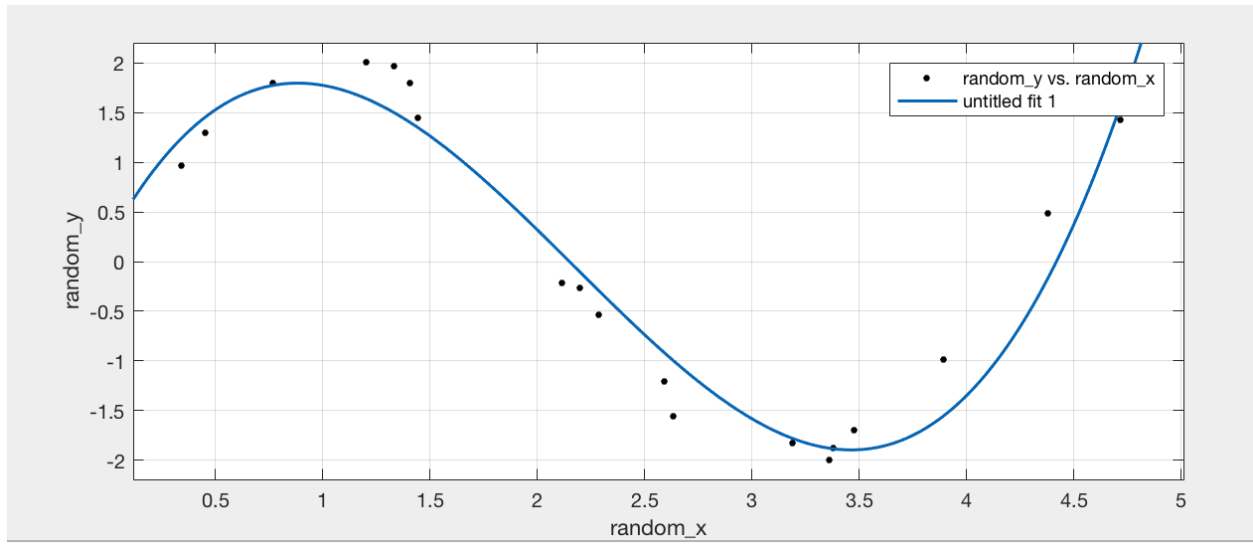
Data training with 1 degree:

SSE: 34.83

R-square 0.1765

Adjusted R-square: 0.1308

RMSE 1.391



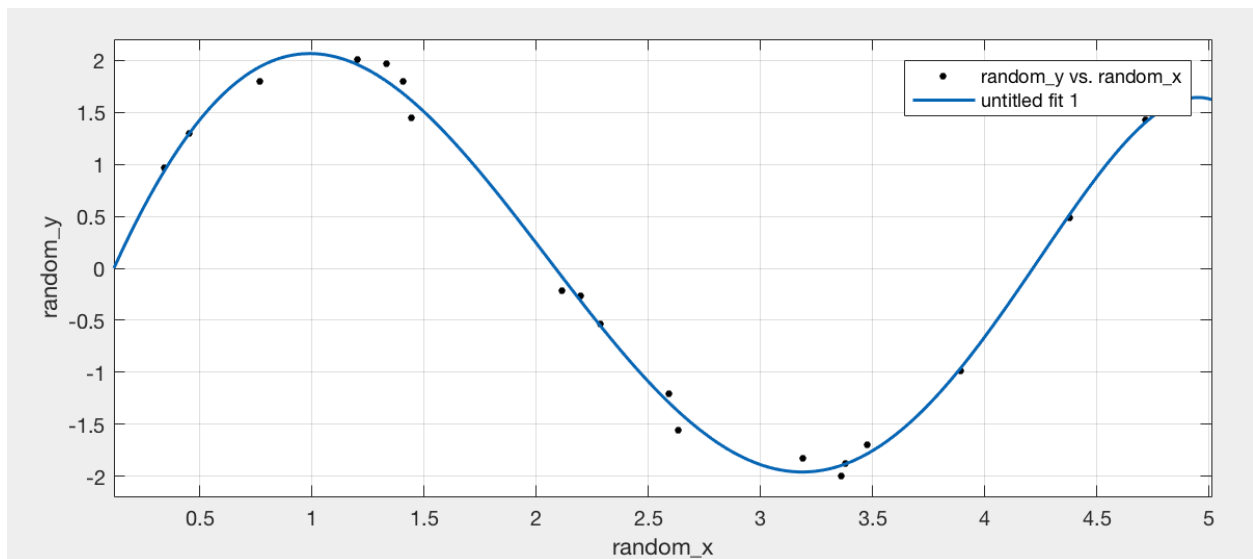
Data Training with Degree 3.

SSE: 2.255

R-square: 0.9467

Adjusted R-square: 0.9367

RMSE: 0.3754



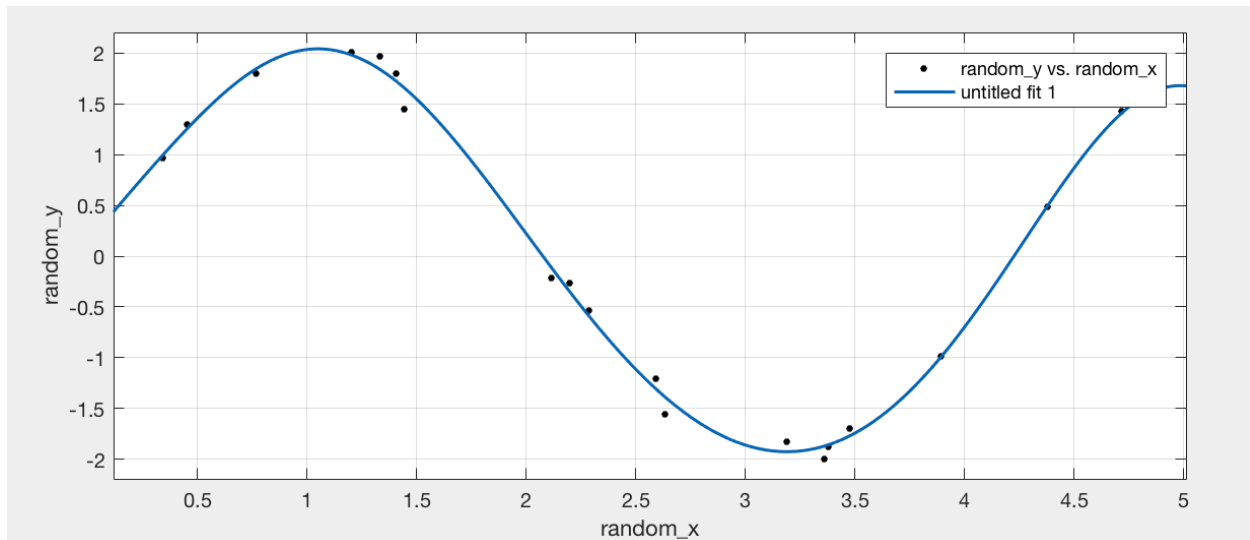
Data Training with Degree 5.

SSE: 0.1852

R-square: 0.9956

Adjusted R-square: 0.9941

RMSE: 0.115



Data Training with Degree 9.

SSE: 0.1571

R-square: 0.9963

Adjusted R-square: 0.9929

RMSE: 0.1254

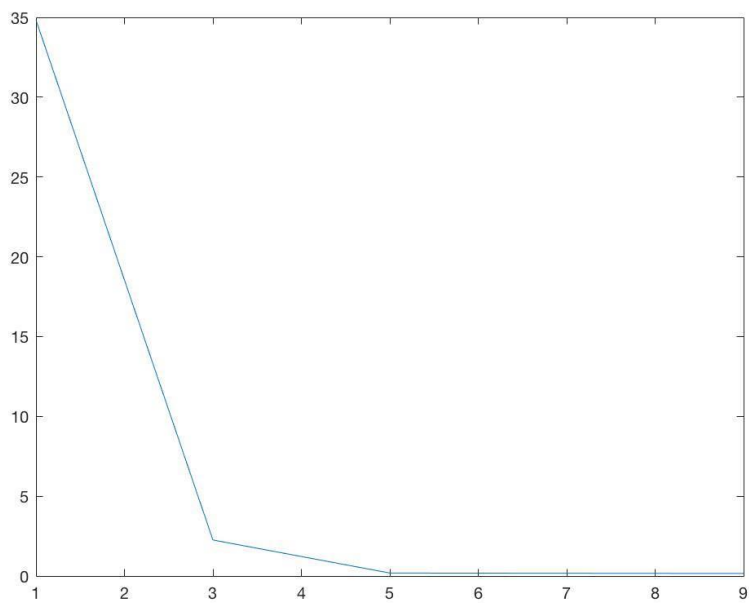
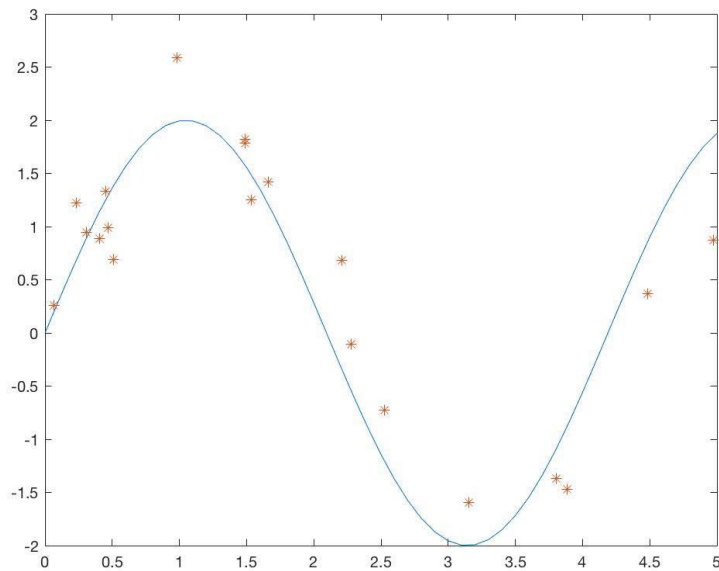
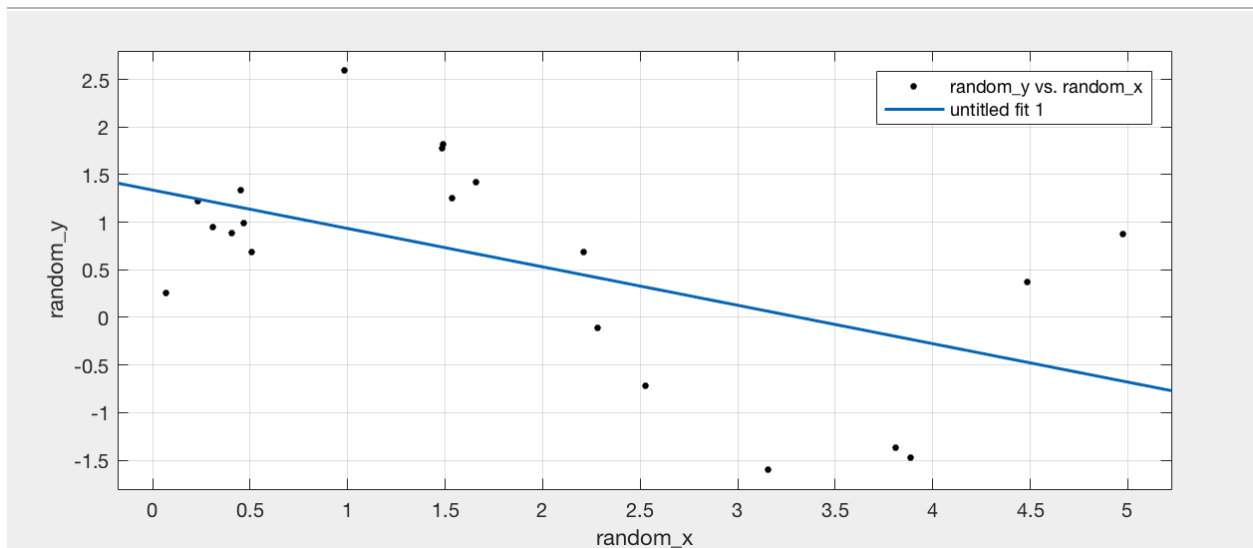


Figure: SSE Value vs Approximation Degree under 0.1 Variance



(Figure:  $y=2\sin(1.5x)$  and 20 random points with 0.1 variance)



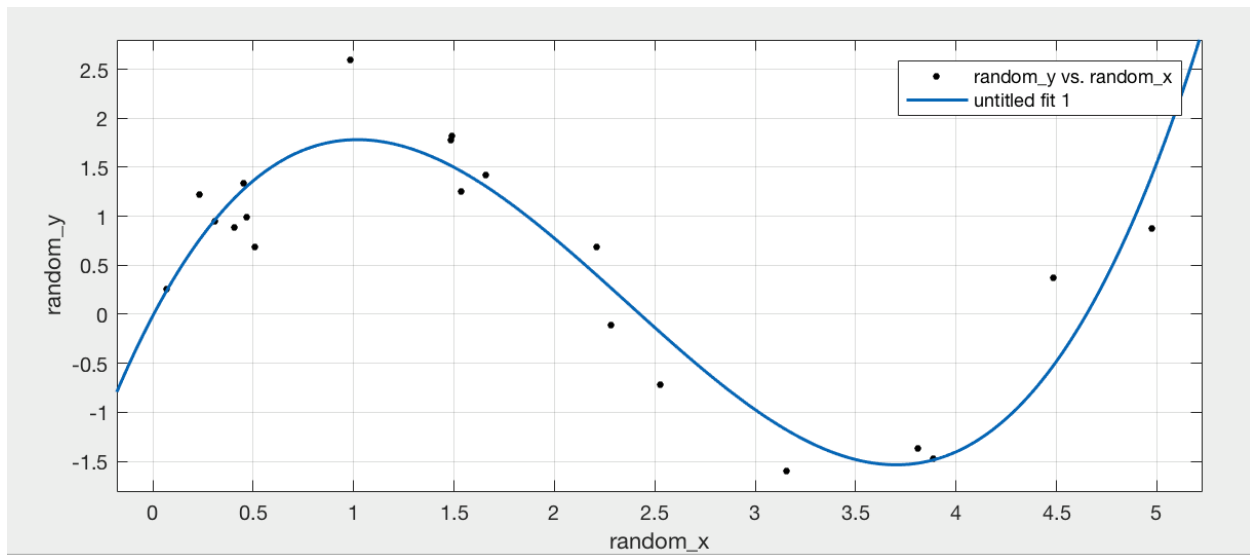
Data training with Degree 1

SSE: 17.52

R-Square: 0.2914

Adjusted R-square: 0.2521

RMSE: 0.9865



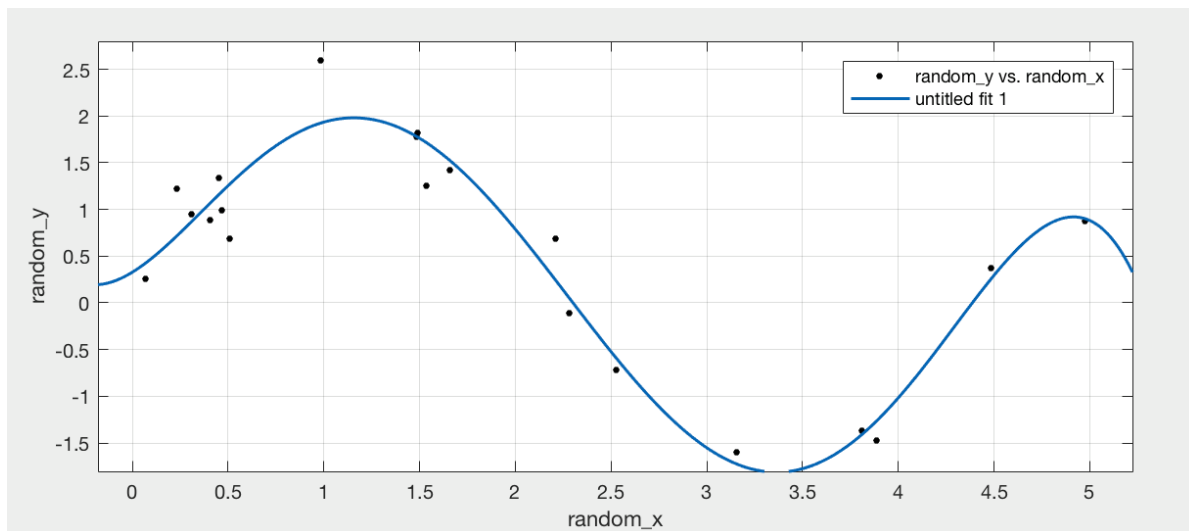
Data training with Degree 3

SSE: 3.558

R-Square: 0.8561

Adjusted R-Square: 0.8291

RMSE: 0.4715



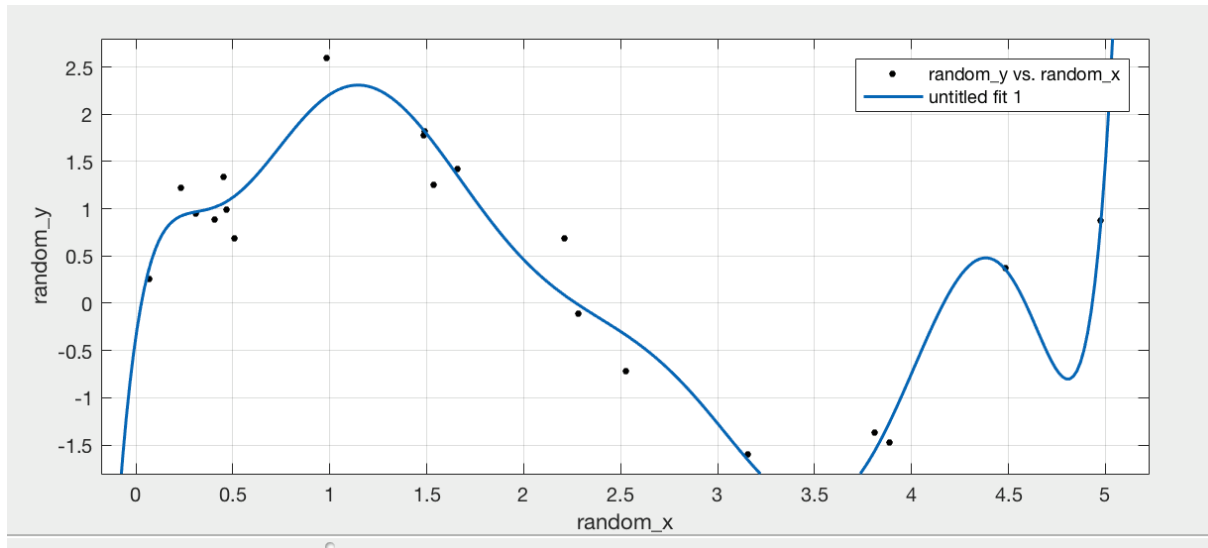
Data training with Degree 5

SSE: 1.708

R-Square: 0.9309

Adjusted R-Square: 0.9062

RMSE: 0.3493



Data training with Degree 9

SSE: 1.371

R-Square: 0.9445

Adjusted R-Square: 0.8946

RMSE: 0.3703

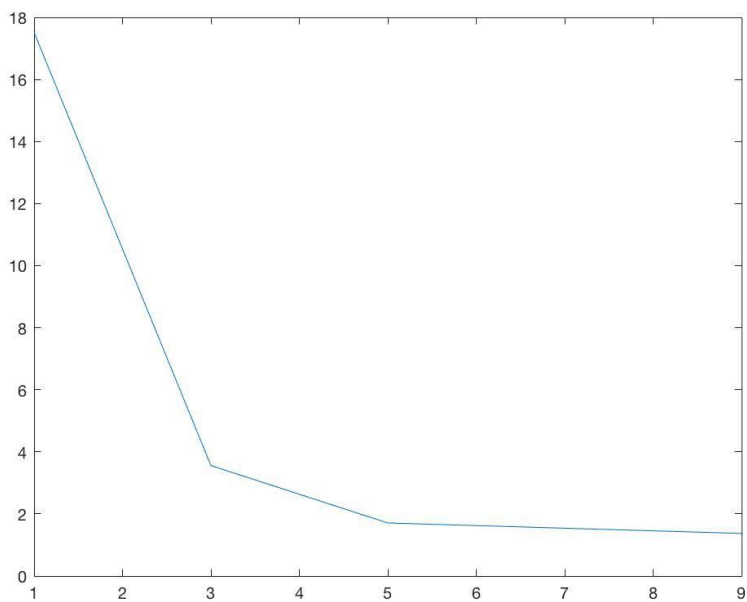


Figure: SSE Value vs Approximation Degree under 0.5 Variance



P2Q4.

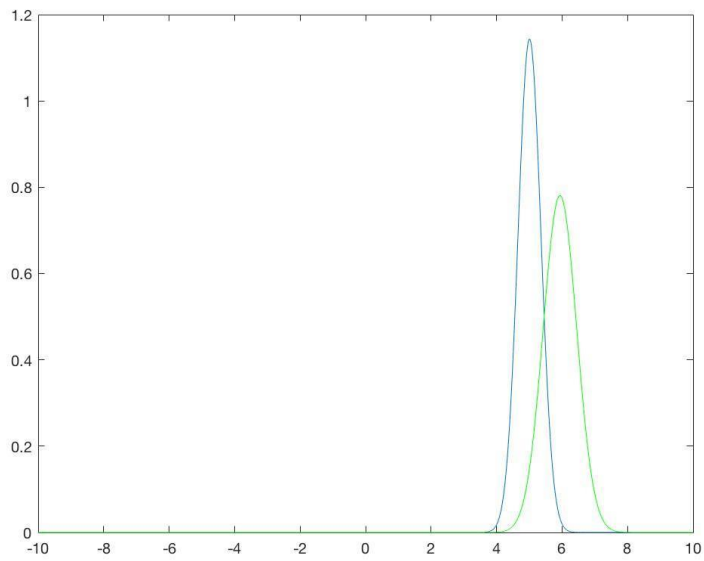


Figure: Maximum Likelihoods of Reduced Iris data.

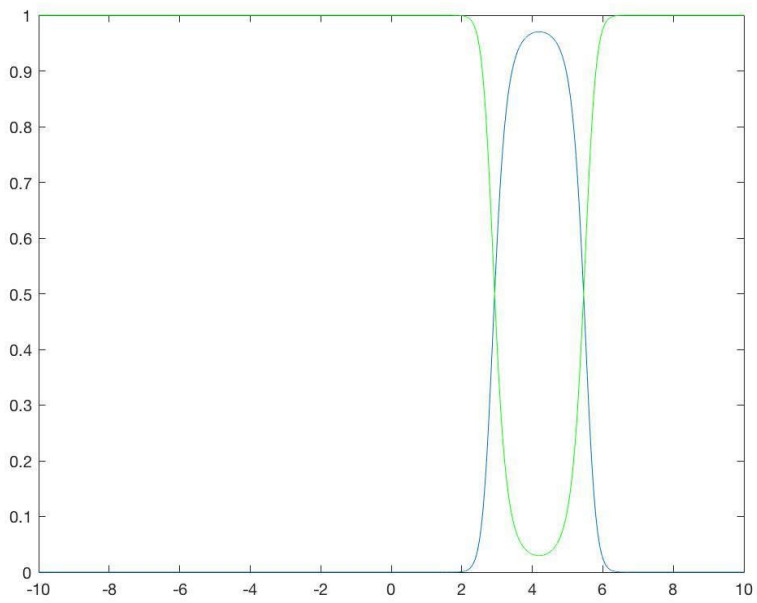


Figure: Class posteriors for Reduced Iris Data

Matlab Code:

```
iris = readtable('iris.txt', 'ReadVariableNames', false);
```

```
iris = table2cell(iris);
```

```
for i = 1:100
```

```
    if i < 51
```

```
        x1(i) = iris(i,1);
```

```
    else
```

```
        x2(i - 50) = iris(i,1);
```

```
    end
```

```
end
```

```
x1 = x1.';
```

```
x2 = x2.';
```

```
x1 = cell2mat(x1);
```

```
x2 = cell2mat(x2);
```

```
mle1 = mle(x1);
```

```
mle2 = mle(x2);
```

```
x_array = [-10:.001:10];
```

```
figure(1);
```

```
y1 = normpdf(x_array, mle1(1), mle1(2));
```

```
plot(x_array, y1)
```

```
hold on
```

```
y2 = normpdf(x_array, mle2(1), mle2(2));
```

```
plot(x_array, y2, 'g')
```

```
hold off
```

```
figure(2);
```

```
for i = 1:20001
    y3(i) = y1(i)/(y1(i)+y2(i));
    y4(i) = y2(i)/(y1(i)+y2(i));
end
plot(x_array, y3)
hold on
plot(x_array, y4, 'g')
```

P3Q1

(a).

Running Matlab code, the number of error we get is 127. Therefore, training classification error is

$$\text{Number of Errors} / \text{Number of Point} = 127 / 500 = 25.4\%$$

**Related Code:**

```
DataSet = readtable('pima_indians_diabetes.csv', 'ReadVariableNames', false);
```

```
DataSet = table2cell(DataSet);
```

```
counter1 = 1;
```

```
counter2 = 1;
```

```
for i = 1:500
```

```
    if (strcmp(DataSet(i,9), 'pos') == 1)
```

```
        c1(counter1,1:8) = DataSet(i,1:8);
```

```
        counter1 = counter1+1;
```

```
    else
```

```
        c2(counter2,1:8) = DataSet(i,1:8);
```

```
        counter2 = counter2+1;
```

```
    end
```

```
end
```

```
valid_interator = 1;
```

```
for i = 1:500
```

```
    valid(valid_interator, 1:8) = DataSet(i,1:8);
```

```
    valid_interator = valid_interator+1;
```

```
end
```

```
c1 = cell2mat(c1);
```

```
c2 = cell2mat(c2);
```

```
valid = cell2mat(valid);
```

```
cov1 = cov(c1);
```

```
cov2 = cov(c2);
```

```
mean1 = mean(c1);
```

```

mean2 = mean(c2);
y1 = mvnpdf(valid, mean1, cov1);
y2 = mvnpdf(valid, mean2, cov2);
error = 0;
predict_value = cell(500,1);
for i = 1:500
    if (y1(i) < y2(i))
        predict_value(i) = cellstr('neg');
    else
        predict_value(i) = cellstr('pos');
    end
    if (strcmp(predict_value(i),DataSet(i,9)) == 0)
        error = error+1;
    end
end
end
(b).

```

Running Matlab code for test classification, the number of error we get is 66. Therefore, test classification error is

$$\text{Number of Errors} / \text{Number of Point} = 66 / 268 = 24.6\%$$

### **Matlab Code**

```

DataSet = readtable('pima_indians_diabetes.csv', 'ReadVariableNames', false);
DataSet = table2cell(DataSet);
valid_iterator = 1;
for i = 501:768
    valid(valid_iterator, 1:8) = DataSet(i,1:8);
    valid_iterator = valid_iterator+1;
end
valid = cell2mat(valid);
z1 = mvnpdf(valid, mean1, cov1);

```

```

z2 = mvnpdf(valid, mean2, cov2);
error = 0;
predict_value = cell(268,1);
for i = 1:268
    if (z1(i) < z2(i))
        predict_value(i) = cellstr('neg');
    else
        predict_value(i) = cellstr('pos');
    end
    if (strcmp(predict_value(i),DataSet(i + 500,9)) == 0)
        error = error+1;
    end
end
end

```

(c).

Mean 1.

4.7802	140.4890	69.7253	21.7143	102.4286	35.3231	0.5672	36.2692
--------	----------	---------	---------	----------	---------	--------	---------

Mean 2.

3.2358	110.0283	67.9748	19.8711	67.5409	29.9921	0.4590	31.1855
--------	----------	---------	---------	---------	---------	--------	---------

Cov1.

8.6603	11.4097	6.1763	-3.9380	-33.8882	1.1000	-0.1219	19.1895
11.4097	806.5008	93.3982	1.7450	1.0654e+03	46.5349	-1.2391	105.4868
6.1763	93.3982	324.8890	53.4794	152.3638	61.4594	-1.2598	42.1466
-3.9380	1.7450	53.4794	218.3146	611.6599	56.4678	0.0052	-23.8814
-33.8882	1.0654e+03	152.3638	611.6599	1.0826e+04	247.3216	7.4645	-164.6149
1.1000	46.5349	61.4594	56.4678	247.3216	65.4896	-0.3639	8.7567
-0.1219	-1.2391	-1.2598	0.0052	7.4645	-0.3639	0.2281	-0.2483
19.1895	105.4868	42.1466	-23.8814	-164.6149	8.7567	-0.2483	139.8740

Cov2.

14.1503	7.6937	10.5415	-4.5880	-29.2644	-3.5087	-0.1273	18.3634
7.6937	968.3728	39.5992	2.7427	1.1351e+03	6.0058	0.7466	60.0168
10.5415	39.5992	494.0015	78.4515	270.3394	7.4423	-0.1845	59.7263
-4.5880	2.7427	78.4515	296.9234	1.2392e+03	35.2939	1.7236	-30.5470
-29.2644	1.1351e+03	270.3394	1.2392e+03	1.9763e+04	30.7702	5.3007	162.4475
-3.5087	6.0058	7.4423	35.2939	30.7702	56.3220	0.3371	-17.0913
-0.1273	0.7466	-0.1845	1.7236	5.3007	0.3371	0.1555	-0.3581
18.3634	60.0168	59.7263	-30.5470	162.4475	-17.0913	-0.3581	114.5514

P3Q2.



**P3Q5.****M and H1 Value:**

-169.769126255560	-169.769126255560	-153.987061905532	-153.987061905532	-	
134.671376212291	-109.769126255560	Inf	-14.6713762122905	Inf	-
74.6713762122905	-109.769126255560	-153.987061905532	-134.671376212291	-	
169.769126255560	-109.769126255560	-109.769126255560	-74.6713762122905	Inf-	
14.6713762122905	-14.6713762122905				

**M and K1 Value:**

694.932323934640	662.732703004150	631.621908886040	601.596535275394
572.653049197447	544.787786823609	517.996948880953	492.276595495771
467.622640256742	444.030843211791	421.496802420375	400.015943563853
379.583506963508	360.194531159384	341.843831951819	324.525975486758
308.235243557651	292.965588778855	278.710576631234	265.463310558606
253.216335267740	241.961512115841	231.689858919299	222.391344655224
214.054627343051	206.666720921067	200.212574287935	194.674543120838
190.031733105538	186.259192689460	183.326935829359	181.198782674900
179.831021952489	179.170927351365	179.155206650030	179.708531501912
180.742389522574	182.154612391893	183.830042752956	185.642865017160
187.461073425890	189.153308990186	190.597821450905	191.692652337506
192.365478282509	192.581196960218	192.345548734539	191.703892562162
190.735427999510	189.544211278481	188.248855405650	186.972716712011
185.835828611042	184.949152576538	184.411131260420	184.306167523733
184.704512294726	185.663057529421	187.226623027147	189.429440923842
192.296647156948	195.845672012320	200.087480427544	205.027650707899
210.667302848118	217.003899481812	224.031947271700	231.743626861874
240.129376977209	249.178453864154	258.879481612308	269.221002376203
280.192028495233	291.782591418975	303.984275732635	316.790721091659
330.198071192257	344.205347611058	358.814727797436	374.031710701231
389.865160061081	406.327223474907	423.433133971073	441.200908766725

459.650966241466	478.805686180156	498.688939763663	519.325614736672
540.741158112279	562.961154338331	586.010951759107	609.915345100391
634.698317081999	660.382838435870	686.990722722914	714.542530399731
743.057515490450	772.553607817864	803.047423870824	834.554299867354

## M and K2 Value

682.398756308608 636.029981737874 592.011872283877 550.321605057753 510.934125197960  
473.822002250326 438.955295775757 406.301434875496 375.825116902666 347.488231217678  
321.249814485490 297.066044766966 274.890282589117 254.673168316788 236.362786411479  
219.904908272007 205.243325725184 192.320285919873 181.077034073381 171.454461765110  
163.393844059325 156.837628322031 151.730212373865 148.018622781879 145.652980651262  
144.586627938675 144.775786318725 146.178633015037 148.753697034641 152.457490261243  
157.241269391662 163.046752757879 169.800473226423 177.406241184336 185.734984166711  
194.611204078624 203.795841702185 212.967145417325 221.705055962300 229.490822413291  
235.739437626780 239.880169709137 241.479787175608 240.366597055864 236.691078191613  
230.885165148683 223.543807053551 215.291976328774 206.688464918496 198.180915465286  
190.100159522958 182.675373515471 176.056102339778 170.333680252830 165.559158728495  
161.757214536351 158.936429487834 157.096539765476 156.233197322903 156.340687776887  
157.412989267352 159.443543215386 162.424116985576 166.343139089153 171.183856170174  
176.922588559044 183.527254379289 190.956211575168 199.157364266154 208.067433625515  
217.611344792461 227.701860417412 238.239896691160 249.116326124965 260.216351201536  
271.427485335865 282.651550002214 293.819801099163 304.908635394184 315.952062987999  
327.047166487992 338.350474896097 350.065927181620 362.427542970892 375.680930513383  
390.067162438721 405.811011585625 423.113950676746 442.151240588265 463.071973600700  
486.000936156639 511.041382660536 538.278095267913 567.780352802729 599.604614209090  
633.796840420339 670.394447667143 709.427920843689 750.922130424768 794.897399579461

### Matlab Code

```
x = [randn(30,1); 5+randn(30,1)];  
  
H = hist(x, 20);  
  
figure(2);  
  
[K1, xi1, bw] = ksdensity(x);  
  
plot(xi1, K1, '--');  
  
hold on;  
  
[K2, xi2] = ksdensity(x, 'width', bw/2);  
  
plot(xi2, K2)  
  
  
M = normpdf(x)  
  
Klmh = sum(M.*log2(M)-log2(H))  
  
Klmk1 = sum(M.*log2(M)-log2(K1))  
  
Klmk2 = sum(M.*log2(M)-log2(K2))
```