

```
In[163]:= (*© 2012-Present Computational ClassNotes,  
lossofgenerality.org, Creative Commons License *)  
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Attribution-NonCommercial-ShareAlike *)
```

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SetOptions[EvaluationNotebook[], Background → White]
```

# Support Vector Regression

Native code, not from Mathematica. Note the maximizer is Differential Evolution.

```

In[28]:= SupportVectorRegression[{xm_, ym_}, K_, e_, c_, method_] := Module[
  {m, n, M, i, j, W, g, vars, sol, b},

  m = Length[ym]; n = Length[xm[[1]]];
  M = Table[K[xm[[i]], xm[[j]]], {i, 1, m}, {j, 1, m}] + 1/c IdentityMatrix[m];
  W =  $\sum_{i=1}^m \text{Subscript}[\alpha, i] ym[[i]] - e \sum_{i=1}^m \text{Abs}[\text{Subscript}[\alpha, i]] -$ 
     $1/2 \sum_{i=1}^m \sum_{j=1}^m (\text{Subscript}[\alpha, i] \text{Subscript}[\alpha, j] M[[i, j]]);$ 
  g = Apply[And, Join[Table[-c < Subscript[\alpha, i] <= c, {i, 1, m}],
    { $\left(\sum_{i=1}^m \text{Subscript}[\alpha, i] == 0\right)$ }]];

  vars = Table[Subscript[\alpha, i], {i, 1, m}];
  sol = NMaximize[{W, g}, vars, Method -> method][[2]];

  b = 1/m Apply[Plus, Table[ $\left(y_m[[j]] - \sum_{i=1}^m \text{Subscript}[\alpha, i] K[xm[[i]], xm[[j]]] -$ 
     $e - \text{Subscript}[\alpha, j]/c\right) /. sol, \{j, 1, m\}]];

  { $\left(\sum_{i=1}^m \text{Subscript}[\alpha, i] K[xm[[i]], \text{Table}[\text{Subscript}[x, j], \{j, 1, n\}]] + b\right) /. sol,$ 
    vars /. sol, M, W, g, vars, sol, b}];

waveletKernel[u_, v_] :=
   $\prod_{i=1}^{\text{Length}[u]} (\text{Cos}[1.75 (u[[i]] - v[[i]])/a] \text{Exp}[-((u[[i]] - v[[i]])^2)/(2 * a^2)]);$ 

rbfKernel[x_, z_] := Exp[-b (x - z).(x - z)];
polyKernel[x_, z_] := (e + x.z)^d;
vovkKernel[x_, z_] := (1 - x.z)^-d;
fourierKernel[x_, z_] :=  $\prod_{i=1}^{\text{Length}[x]} (1 - q^2) / (2 (1 - 2 q \text{Cos}[x[[i]] - z[[i]]] + q^2));$$ 
```

## Data

```

In[8]:= testdata = Import[
  FileNameJoin[{NotebookDirectory[], "aggregatefeaturesConorTest.csv"}], "CSV"];
data = Import[FileNameJoin[{NotebookDirectory[],
  "aggregatefeaturesConorTrain.csv"}], "CSV"];
Length@data
Length@testdata

Out[10]= 68

Out[11]= 67

```

## Regression

```

In[119]:=
input = Map[Drop[#, 1] &, data];
output = 0.3 * Map[First@# &, data];
testinput = Map[Drop[#, 1] &, testdata];
testoutput = 0.3 * Map[First@# &, testdata];

(*some experimentation needed to find the optimum values,
but we could use another search optimized on top to do just that*)
epsilon = 0.0025; c = 200.0; a = 14;

method = "DifferentialEvolution";

(*there are many more kernels, there is an entire theory for kernels in math.*)
kernel = waveletKernel;

(*Note {input,output} was used corresponding to training data*)
F = SupportVectorRegression[{input, output}, kernel, epsilon, c, method];

```

## Quadratic Kernel Regression

```
In[158]:= Short[F[[1]], 20]
```

```
Out[158]/Short= 99.4263 - 112.44
```

$$e^{-\frac{x_1^2}{392} - \frac{x_2^2}{392} - \frac{1}{392} (0.17126 - x_3)^2 - \frac{1}{392} (0.79574 - x_4)^2 - \frac{1}{392} (0.1713 - x_5)^2 - \frac{1}{392} (1.937 - x_6)^2 - \frac{1}{392} (5.2538 - x_7)^2 - \frac{1}{392} (9.5958 - x_8)^2 - \frac{1}{392} (2.9 - x_9)^2} \cos[0.125 x_1] \cos[0.125 x_2] \cos[0.125 (0.17126 - x_3)] \cos[0.125 (0.79574 - x_4)] \cos[0.125 (0.1713 - x_5)] \cos[0.125 (1.937 - x_6)] \cos[0.125 (5.2538 - x_7)] \cos[0.125 (9.5958 - x_8)] \cos[0.125 (2.9 - x_9)] + \ll 97 \gg + 47.6064$$

$$e^{-\frac{x_1^2}{392} - \frac{1}{392} (3.8839 - x_2)^2 - \frac{1}{392} (11.141 - x_3)^2 - \frac{1}{392} (7.9125 - x_4)^2 - \frac{1}{392} (8.64 - x_5)^2 - \frac{1}{392} (12.952 - x_6)^2 - \frac{1}{392} (20.093 - x_7)^2 - \frac{1}{392} (44.913 - x_8)^2 - \frac{1}{392} (22.2 - x_9)^2} \cos[0.125 x_1] \cos[0.125 (3.8839 - x_2)] \cos[0.125 (11.141 - x_3)] \cos[0.125 (7.9125 - x_4)] \cos[0.125 (8.64 - x_5)] \cos[0.125 (12.952 - x_6)] \cos[0.125 (20.093 - x_7)] \cos[0.125 (44.913 - x_8)] \cos[0.125 (22.2 - x_9)]$$

```
In[51]:= (*https://en.wikipedia.org/wiki/Coefficient_of_determination*)
```

```
rsQUARED[y_, yhat_] := Module[{ybar, SStot, SSreg, SSres},
```

```
  ybar = Mean@y;
```

```
  SStot = Total@Map[(# - ybar) ^ 2 &, y];
```

```
  SSreg = Total@Map[(# - ybar) ^ 2 &, yhat];
```

```
  SSres = Total@Table[(y[[i]] - yhat[[i]]) ^ 2, {i, 1, Length@y}];
```

```
  1 - (SSres / SStot)
```

```
];
```

```
In[159]:= rsQUARED[testoutput ,
```

```
  Table[F[[1]] /. (Flatten@Table[{Subscript[x, j] -> testinput[[i]][[j]]},  
    {j, 1, Length@testinput}]]), {i, 1, Length@testinput}]]
```

```
Out[159]= 0.935999
```

## SVR

```
In[164]:= svr = Table[F[[1]] /. (Flatten@Table[{Subscript[x, j] -> testinput[[i]][[j]]},  
  {j, 1, Length@testinput}]]), {i, 1, Length@testinput}]
```

```
Out[164]= {95.2982, 98.8733, 92.6768, 97.4578, 90.7424, 107.898, 97.3888, 104.944, 35.5161,  
  131.295, 149.335, 136.621, 148.474, 112.923, 147.173, 111.47, 143.887,  
  30.4932, 117.881, 120.452, 114.871, 120.219, 104.723, 128.409, 103.096,  
  129.83, 44.9977, 126.383, 124.31, 126.947, 124.701, 104.892, 98.4552, 104.254,  
  101.502, 23.6242, 115.982, 110.205, 118.898, 106.302, 159.607, 126.483,  
  149.391, 131.653, 39.7022, 155.437, 157.84, 155.116, 154.113, 131.377, 158.154,  
  135.731, 160.141, 42.7113, 151.746, 155.392, 152.956, 155.941, 143.558,  
  155.606, 141.267, 159.424, 34.7847, 157.023, 154.433, 157.665, 155.969}
```

# Test Data

In[161]:=

**testoutput**

Out[161]= { 105.285, 98.487, 106.056, 97.68, 85.692, 100.008, 84.093, 101.265, 25.4403, 144.399, 150.855, 144.438, 151.266, 134.718, 147.012, 132.93, 145.314, 36.903, 118.689, 128.703, 115.053, 127.077, 114.741, 125.388, 115.767, 129.663, 31.392, 119.136, 127.644, 117.321, 127.881, 95.328, 106.266, 92.811, 106.377, 24.1647, 110.442, 107.679, 110.418, 107.643, 136.776, 156.867, 134.88, 153.858, 42.519, 148.422, 160.779, 152.295, 160.548, 144.609, 163.824, 143.43, 162.9, 42.933, 163.164, 163.953, 161.004, 164.436, 143.016, 153.912, 143.871, 155.097, 35.85, 164.649, 156.477, 163.029, 158.91 }

In[167]:=

**ListLinePlot[{svr, testoutput}, PlotLegends → {"SVR", "Test Data"}]**

