#### **Regularization with Weight Decay**

In the following experiments use the data provided in the files **in.dta**, **out.dta** as a training and test set respectively. Each line of the files corresponds to a two-dimensional input  $x = (x_1, x_2)$ , so that  $X = \mathbb{R}^2$ , followed by the corresponding label from  $\mathcal{Y} = \{-1, 1\}$ . We are going to apply Linear Regression with a non-linear transformation for classification. The nonlinear transformation is given by

$$\varphi(x_1, x_2) = (1, x_1, x_2, x_1^2, x_2^2, x_1 x_2, |x_1 - x_2|, |x_1 + x_2|).$$

Recall that the classification error is defined as the fraction of misclassified points.

```
(* Clear globals *)
Clear[GenerateX, DoLinearRegressionExperiment, ClassificationE]
```

```
GenerateX[OptionsPattern[]] :=
 Module {
    \mathcal{D}y,\,\mathcal{D},\,y,\,f\,,\,d=2\,,
    f1, f2,
    m, a, b, c,
    t1, t2,
    dotTest1, dotTest2, DotTest
   f1 = RandomReal[{-1, 1}, {1, d}][[1]];
   f2 = RandomReal[{-1, 1}, {1, d}][[1]];
        f1[[2]] - f2[[2]]
f1[[1]] - f2[[1]];
  a = -m;
  b = 1;
  c = m f1[[1]] - f1[[2]];
  f = \{\{c\}, \{a\}, \{b\}\};
   (* f should not dot to zero for our two original points! *)
  t1 = {1, f1[[1]], f1[[2]]};
  t2 = {1, f2[[1]], f2[[2]]};
  DotTest[v_, s_] := If[Abs[v] > 0.0000000001, Throw[s], 0];
  DotTest[t1.f, "t1 dot test failed"];
  DotTest[t2.f, "t2 dot test failed"];
   (*D = RandomReal[\{-1,1\}, \{OptionValue[DSize], d\}]; *)
  Dy = Import[OptionValue[R2DataFile], "Table"];
  \mathcal{D} = \mathcal{D}y \, [\, [\, \textbf{All} \, (\, \star \, \textbf{1} \, ; \, ; \, \textbf{15} \, \star \,) \, \, , \, \, \{\, \textbf{1} \, , \, \, d\, \} \, ] \, ] \, ;
  y = \mathcal{D}y[[All(*1;;15*), d+1]];
  \{\mathcal{D}, f, y\}
Options[GenerateX] = {R2DataFile → "in.dta"};
```

```
LinearTarget[f_{-}, X_{-}] := Sign[X.f];
NoFeature[X_] := X;
\phi[X_] := \{1, \#2, \#3, \#2^2, \#3^2, \#2 \#3, Abs[\#2 - \#3], Abs[\#2 + \#3]\} \& @@@X;
NoRegularizer[X_{-}, \lambda_{-}] := PseudoInverse[X_{-}, \lambda_{-}] := VseudoInverse[X_{-}, \lambda_{-}] :=
Tikhonov[X_{-}, \lambda_{-}] := Inverse[X^{T}.X + \lambda IdentityMatrix[Dimensions[X][[2]]]].X^{T};
DoLinearRegressionExperiment[X , OptionsPattern[]] :=
   Module[{
           \mathcal{D}, f, y, y,
          X, Xf, Xfdag,
          w
       },
       \{\mathcal{D}, f, y\} = X;
       X = Function[x, Prepend[x, 1]] /@D;
       Xf = OptionValue[DataFeature][X];
       Xfdag = OptionValue[Regularizer][Xf, OptionValue[<math>\lambda]];
       w = Xfdag.y;
       \{w, X, y, D, f\}
Options[DoLinearRegressionExperiment] = {TargetFunction → LinearTarget,
          DataFeature \rightarrow \phi, Regularizer \rightarrow NoRegularizer, \lambda \rightarrow 0;
ClassificationE[X_, w_, y_, OptionsPattern[]] :=
   Determine classification error for hypothesis w when
           given data collection X and labels y. NOTE that this is a
           generic classification error function that is E_{in}/E_{out}-agnostic
   *)
   Module {
          N, misses, sumOfMisses
       },
       N = Length[y];
       misses =
          MapThread[If[#1 # #2, 1, 0] &, {Sign[OptionValue[DataFeature][X].w], y}];
       sumOfMisses = Total[misses];
              sumOfMisses
        N
Options[ClassificationE] = {DataFeature \rightarrow \phi};
```

## Linear regression with non-linear transform

Run Linear Regression on the training set after performing the non-linear transformation to determine  $E_{\rm in}$  and  $E_{\rm out}$ .

```
(* Clear globals *)
Clear[Experiment1, elEin, elEout]
```

```
Experiment1[] :=
 (*
 Perform linear regression using the provided in-sample file,
 then determine \mathbf{E}_{\text{in}} from it and \mathbf{E}_{\text{out}} from the provided out-
  of-sample data+labelage.
 *)
 Module[{
   Ein, Eout,
   Din, Dout,
   Xin, Xout,
   yin, yout,
   \mathbf{w}, f, d = 2
   \{w, Xin, yin, Din, f\} =
   DoLinearRegressionExperiment[GenerateX[R2DataFile → "in.dta"]];
  Ein = ClassificationE[Xin, w, yin];
  \{Dout, f, yout\} = Generate X[R2DataFile \rightarrow "out.dta"];
  Xout = Function[x, Prepend[x, 1]] /@ Dout;
  Eout = ClassificationE[Xout, w, yout];
  {Ein, Eout}
 ]
{elEin, elEout} = Experiment1[];
StringForm[
 "Linear regression results:\nE<sub>in</sub>=``, E<sub>out</sub>=``", elEin × 1., elEout × 1.]
Linear regression results:
E_{in}=0.02857142857142857, E_{out}=0.084
```

## Linear regression with weight decay, k=-3

Now add weight decay to Linear Regression, that is, add the term  $\frac{\lambda}{N} \sum_{i=0}^{7} w_i^2$  to the squared in-sample error, using  $\lambda = 10^k$ , k = -3.

```
(* Clear globals *)
Clear[Experiment2, e2Ein, e2Eout, e2\lambda]
```

```
Experiment2[regularizer\lambda_] :=
  (*
 Perform linear regression using the provided in-sample file,
 then determine Ein from it and Eout from the provided out-
  of-sample data+labelage.
 *)
 Module[{
    Ein, Eout,
    Din, Dout,
    Xin, Xout,
    yin, yout,
    w, f, d = 2
   \{w, Xin, yin, Din, f\} = DoLinearRegressionExperiment[
     GenerateX [\mathbb{R}2DataFile \rightarrow "in.dta"], Regularizer \rightarrow Tikhonov, \lambda \rightarrow regularizer\lambda];
   Ein = ClassificationE[Xin, w, yin];
   \{Dout, f, yout\} = Generate X[R2DataFile \rightarrow "out.dta"];
   Xout = Function[x, Prepend[x, 1]] /@ Dout;
  Eout = ClassificationE[Xout, w, yout];
   {Ein, Eout}
 1
e2\lambda = 10^{-3};
{e2Ein, e2Eout} = Experiment2[e2λ];
StringForm[
 "Linear regression results (w/ \lambda=`` regularization):\nE<sub>in</sub>=``, E<sub>out</sub>=``",
 e2\lambda, e2Ein \times 1., e2Eout \times 1.]
Linear regression results (w/ \lambda = \frac{1}{1000} regularization):
E_{in} = 0.02857142857142857, E_{out} = 0.08
```

# Linear regression with weight decay, k=3

(\* Clear globals \*)

```
Clear[e3Ein, e3Eout, e3λ]
e3\lambda = 10^3;
\{e3Ein, e3Eout\} = Experiment2[e3\lambda];
StringForm[
 "Linear regression results (w/ \lambda=`` regularization):\nE<sub>in</sub>=``, E<sub>out</sub>=`",
 e3\lambda, e3Ein \times 1., e3Eout \times 1.]
Linear regression results (w/ \lambda=1000 regularization):
```

## Linear regression with weight decay, various k

 $E_{in} = 0.37142857142857144$ ,  $E_{out} = 0.436$ 

Looking for value of k that achieves the smallest out-of-sample classification error, trying integer k from -2 to 2.

```
(* Clear globals *)
 Clear[e4Ein, e4Eout, e4λ, i]
e4\lambda = \{10^2, 10^1, 10^0, 10^{-1}, 10^{-2}\};
i = 1;
{e4Ein, e4Eout} = Experiment2[e4λ[[i]]];
StringForm[
 "Linear regression results (w/ \lambda=`` regularization):\nE<sub>in</sub>=``, E<sub>out</sub>=`",
 e4\lambda[[i]], e4Ein \times 1., e4Eout \times 1.]
i = 2;
{e4Ein, e4Eout} = Experiment2[e4λ[[i]]];
StringForm[
 "Linear regression results (w/ \lambda=`` regularization):\nE<sub>in</sub>=``, E<sub>out</sub>=``",
 e4\lambda[[i]], e4Ein \times 1., e4Eout \times 1.]
\{e4Ein, e4Eout\} = Experiment2[e4\lambda[[i]]];
StringForm[
 "Linear regression results (w/ \lambda=`` regularization):\nE<sub>in</sub>=``, E<sub>out</sub>=``",
 e4\lambda[[i]], e4Ein \times 1., e4Eout \times 1.]
i = 4;
{e4Ein, e4Eout} = Experiment2[e4λ[[i]]];
StringForm[
 "Linear regression results (w/ \lambda=`` regularization):\nE<sub>in</sub>=``, E<sub>out</sub>=``",
 e4\lambda[[i]], e4Ein \times 1., e4Eout \times 1.]
i = 5;
{e4Ein, e4Eout} = Experiment2[e4λ[[i]]];
 "Linear regression results (w/ \lambda=`` regularization):\nE<sub>in</sub>=``, E<sub>out</sub>=`",
 e4\lambda[[i]], e4Ein \times 1., e4Eout \times 1.]
Linear regression results (w/ \lambda=100 regularization):
E_{in}=0.2, E_{out}=0.228
Linear regression results (w/ \lambda=10 regularization):
E_{in} = 0.05714285714285714, E_{out} = 0.124
Linear regression results (w/ \lambda=1 regularization):
E_{in}=0., E_{out}=0.092
Linear regression results (w/ \lambda = \frac{1}{10} regularization):
E_{in}=0.02857142857142857, E_{out}=0.056
Linear regression results (w/ \lambda = \frac{1}{100} regularization):
E_{in}=0.02857142857142857, E_{out}=0.084
```