

## Bias and Variance

Consider the case where the target function  $f: [-1, 1] \Rightarrow \mathbb{R}$  is given by  $f(x) = \sin(\pi x)$  and the input probability distribution is uniform on  $[-1, 1]$ . Assume that the training set has only two examples (picked independently), and that the learning algorithm produces the hypothesis that minimizes the mean squared error on the examples.

```
(* Clear globals *)
Clear[GenerateX, DoLinearRegressionExperiment, Ein,
  GetYInterceptForm, LinearTarget, NoFeature, RiskNoiseFlip]
Clear[DoInteractiveLR, DoAutoLR]

(* Generate our X=[-1,1]x[-1,1] space with d uniformly distributed points
  (input to f) and a random X-partitioning solution function f. *)

GenerateX[OptionsPattern[]] :=
Module[{
  D, f
},
  D = RandomReal[{-1, 1}, OptionValue[DSize]];
  f = OptionValue[Targetf] /@ D; (* (Tanh[ $\pi$  #]) & /@ D; *)
  {D, f}
]

Options[GenerateX] = {DSize -> 2, Targetf -> ((Sin[ $\pi$  #]) &)};

RegressionTarget[f_, X_] := f; (* Sign[X.f]; *)
NoFeature[X_] := X;

RiskNoiseFlip[percent_] := If[RandomReal[] <  $\frac{\text{percent}}{100}$ , -1, 1];

DoLinearRegressionExperiment[X_, OptionsPattern[]] :=
Module[{
  D, f, y,
  X, Xf, Xfdag,
  w
},
  {D, f} = X;
  X = ({0, #}) & /@ D; (* Function[x, Prepend[x, (*1*) 0]] /@ D; *)
  Xf = OptionValue[DataFeature][X];
  y = RiskNoiseFlip[OptionValue[Noise]] * OptionValue[TargetFunction][f, X];
  Xfdag = PseudoInverse[Xf];
  w = Xfdag.y;
  {w, X, y, D, f}
]

Options[DoLinearRegressionExperiment] =
  {TargetFunction -> RegressionTarget, DataFeature -> NoFeature, Noise -> 0};

RegressionsEin[X_, w_, y_] :=
Module[{
  N
},
  N
```

```

N = Length[Y];
1
- Norm[Sign[X.w] - Y]^2
N
]

(*
AltEin[X_,w_,Y_] :=
Module[{
  N
},
N=Length[Y];
1
N Sum_{n=1}^N (Transpose[w].X[[n]]-Y[[n]])^2
]
*)

GetYInterceptForm[w_] :=
Module[{
  a, b, c, M, B
},
a = w[[2]];
b = w[[3]];
c = w[[1]];
M = -a/b; (*If[b!=0,-a/b,10000];*)
B = -c/b; (*If[b!=0,-c/b,10000];*)
{M, B}
]

(* Auto mode implements what the problem asks for:
it runs the tests and then outputs a fomatted string bearing the results. *)
DoAutoLR[OptionsPattern[]] :=
Module[{
  runs = 10 000, N = 100,
  g, gs,
  Y, Ys,
  X, Xs,
  f, fs,
  D,
  avg,
  (*avEin,*)
  i, n
},
fs = {};
gs = {};
Ys = {};
Xs = {};
For[i = 1, i <= runs, i++, (
  {g, X, Y, D, f} = DoLinearRegressionExperiment[
    GenerateX[DSize -> 2(*,Targetf->((Tanh[pi#])&)*)],

```

```

      DataFeature -> OptionValue[DataFeature] (*, OptionValue[DataFeature] *)];
    AppendTo[fs, f];
    AppendTo[gs, g];
    AppendTo[ys, y];
    AppendTo[Xs, X]
  ]];
  avg = {Mean[gs[[All, 1]]], Mean[gs[[All, 2]]]};
  (*avEin=
    Mean[MapThread[Function[{x,y,z}, ClassificationEin[x,y,z]], {Xs,gs,ys}]]];*)
  n = Length[D];
  {fs, gs, D, avg, runs, n}
]
Options[DoAutoLR] = {DataFeature -> NoFeature};

```

### Finding the expected value

Assume the learning model consists of all hypotheses of the form  $h(x) = ax$ , find the expected value  $\bar{g}(x)$  of the hypothesis produced by the learning algorithm (expected value with respect to the data set).

```

(* Clear globals *)
Clear[Experiment1]
Clear[elavg, elfs, elgs, elD, elruns, eln]

```

```

Experiment1[interactive_] := If[interactive,
  DoAutoLR[DataFeature -> NoFeature], DoAutoLR[DataFeature -> NoFeature]]
{elfs, elgs, elD, elavg, elruns, eln} = Experiment1[False];
(*Experiment1[True]*)
StringForm[
  "Linear regression results\n(runs=`, N=`):\naverage g offset=``\naverage
  g coefficient=``", elruns, eln, elavg[[1]] × 1., elavg[[2]] × 1.]
Linear regression results
(runs=10000, N=2):
average g offset=6.044187200081877`*^-19
average g coefficient=1.4468213610288414`

```

### Finding the bias

```

(* Clear globals *)
Clear[Experiment2]
Clear[bias]

```

```

Experiment2[gbar_] :=
Module[{bias},
  bias =  $\frac{1}{2} \int_{-1}^1 (\text{gbar}[[1]] + \text{gbar}[[2]] * x - \text{Sin}[\pi x])^2 dx$ 
  (*bias:= $\frac{1}{2}$ Total[MapThread[(gbar*#1-#2)^2&,{xs,fs}]]*)
]
bias = Experiment2[elavg];
StringForm["Linear regression results\nbias=`", bias*1.]
Linear regression results
bias=0.2766889313950476`

```

### Finding the variance

```

(* Clear globals *)
Clear[Experiment3]
Clear[variance]

Experiment3[gbar_, gs_] :=
Module[{variance},
  variance =  $\left( \frac{1}{\text{Length}[gs]} \sum_{i=1}^{\text{Length}[gs]} \left( \frac{1}{2} \int_{-1}^1 ((gs[[i, 1]] + gs[[i, 2]] * x) - (\text{gbar}[[1]] + \text{gbar}[[2]] * x))^2 dx \right) \right)$ 
  (*bias:= $\frac{1}{2}$ Total[MapThread[(gbar*#1-#2)^2&,{xs,fs}]]*)
]
variance = Experiment3[elavg, elgs];
StringForm["Linear regression results\nvariance=`", variance*1.]
Linear regression results
variance=0.2392776493640879`

```

### Check expected value of $E_{\text{out}}$ for different learning models

Now change  $\mathcal{H}$ . Check calculated expected value for out-of-sample error for several learning models

```

(* Clear globals *)
Clear[Experiment4]
Clear[e4avg, e4fs, e4gs, e4D, e4runs, e4n, e4EdEoutGd]

```

```

FeatureA[X_] := {1, 0} & @@@ X;
FeatureB[X_] := {0, #2} & @@@ X;
FeatureC[X_] := {1, #2} & @@@ X;
FeatureD[X_] := {0, #2^2} & @@@ X;
FeatureE[X_] := {1, #2^2} & @@@ X;

{e4fs, e4gs, e4D, e4avg, e4runs, e4n} = DoAutoLR[DataFeature → FeatureA];
bias = Experiment2[e4avg];
variance = Experiment3[e4avg, e4gs];
e4EdEoutGd = bias + variance;
StringForm["Linear regression results h(x)= b\n(runs=`,
    N=``):\naverage g offset=``\naverage g coefficient=``\nED[Eout(gD)] = ``",
    e4runs, e4n, e4avg[[1]] × 1., e4avg[[2]] × 1., e4EdEoutGd × 1.]

{e4fs, e4gs, e4D, e4avg, e4runs, e4n} = DoAutoLR[DataFeature → FeatureB];
bias = Experiment2[e4avg];
variance = Experiment3[e4avg, e4gs];
e4EdEoutGd = bias + variance;
StringForm["Linear regression results h(x)= ax\n(runs=`,
    N=``):\naverage g offset=``\naverage g coefficient=``\nED[Eout(gD)] = ``",
    e4runs, e4n, e4avg[[1]] × 1., e4avg[[2]] × 1., e4EdEoutGd × 1.]

{e4fs, e4gs, e4D, e4avg, e4runs, e4n} = DoAutoLR[DataFeature → FeatureC];
bias = Experiment2[e4avg];
variance = Experiment3[e4avg, e4gs];
e4EdEoutGd = bias + variance;
StringForm["Linear regression results h(x)= ax + b\n(runs=`,
    N=``):\naverage g offset=``\naverage g coefficient=``\nED[Eout(gD)] = ``",
    e4runs, e4n, e4avg[[1]] × 1., e4avg[[2]] × 1., e4EdEoutGd × 1.]

{e4fs, e4gs, e4D, e4avg, e4runs, e4n} = DoAutoLR[DataFeature → FeatureD];
bias = Experiment2[e4avg];
variance = Experiment3[e4avg, e4gs];
e4EdEoutGd = bias + variance;
StringForm["Linear regression results h(x)= ax^2\n(runs=`,
    N=``):\naverage g offset=``\naverage g coefficient=``\nED[Eout(gD)] = ``",
    e4runs, e4n, e4avg[[1]] × 1., e4avg[[2]] × 1., e4EdEoutGd × 1.]

{e4fs, e4gs, e4D, e4avg, e4runs, e4n} = DoAutoLR[DataFeature → FeatureE];
bias = Experiment2[e4avg];
variance = Experiment3[e4avg, e4gs];
e4EdEoutGd = bias + variance;
StringForm["Linear regression results h(x)= ax^2 + b\n(runs=`,
    N=``):\naverage g offset=``\naverage g coefficient=``\nED[Eout(gD)] = ``",
    e4runs, e4n, e4avg[[1]] × 1., e4avg[[2]] × 1., e4EdEoutGd × 1.]

```

```

Linear regression results  $h(x) = b$ 
(runs=10000, N=2):
average  $\mathcal{G}$  offset=0.010487445035393929`
average  $\mathcal{G}$  coefficient=0.`
 $\mathbb{E}_{\mathcal{D}}[\mathbb{E}_{\text{out}}(\mathcal{G}^{\mathcal{D}})] = 0.7542300448857159`$ 

Linear regression results  $h(x) = ax$ 
(runs=10000, N=2):
average  $\mathcal{G}$  offset= $-1.46478 \times 10^{-19}$ 
average  $\mathcal{G}$  coefficient=1.4279040025903553`
 $\mathbb{E}_{\mathcal{D}}[\mathbb{E}_{\text{out}}(\mathcal{G}^{\mathcal{D}})] = 0.5079414245539238`$ 

Linear regression results  $h(x) = ax + b$ 
(runs=10000, N=2):
average  $\mathcal{G}$  offset=0.0028060964029559825`
average  $\mathcal{G}$  coefficient=0.7802982730255525`
 $\mathbb{E}_{\mathcal{D}}[\mathbb{E}_{\text{out}}(\mathcal{G}^{\mathcal{D}})] = 1.8500665718437155`$ 

Linear regression results  $h(x) = ax^2$ 
(runs=10000, N=2):
average  $\mathcal{G}$  offset=1.5827938841056742`* $^{-18}$ 
average  $\mathcal{G}$  coefficient=-0.0814262
 $\mathbb{E}_{\mathcal{D}}[\mathbb{E}_{\text{out}}(\mathcal{G}^{\mathcal{D}})] = 16.4040487923253`$ 

Linear regression results  $h(x) = ax^2 + b$ 
(runs=10000, N=2):
average  $\mathcal{G}$  offset=0.4649135588013267`
average  $\mathcal{G}$  coefficient=-0.553618
 $\mathbb{E}_{\mathcal{D}}[\mathbb{E}_{\text{out}}(\mathcal{G}^{\mathcal{D}})] = 6584.2916889760345`$ 

```