

10-601
Machine Learning

Problem Set 2

MATLAB tips

- Conventional wisdom: MATLAB *hates* loops
 - May be less of an issue with most recent versions
 - Ideally use matrix operations whenever possible
- Examples:

```
A =
```

```
    1    2    3  
    4    5    6
```

MATLAB tips

A =			
	1	2	3
	4	5	6

- Conventional wisdom: MATLAB *hates* loops
 - May be less of an issue with most recent versions
 - Ideally use matrix operations whenever possible
- Examples:

```
>> exp(A)
```

```
ans =
```

```
    2.7183    7.3891   20.0855  
   54.5982  148.4132  403.4288
```

MATLAB tips

A =			
	1	2	3
	4	5	6

- Conventional wisdom: MATLAB *hates* loops
 - May be less of an issue with most recent versions
 - Ideally use matrix operations whenever possible
- Examples:

```
>> 10 + A
```

```
ans =
```

```
11    12    13
14    15    16
```

```
>> 2 ./ A
```

```
ans =
```

```
2.0000    1.0000    0.6667
0.5000    0.4000    0.3333
```

MATLAB tips

A =			
	1	2	3
	4	5	6

- Conventional wisdom: MATLAB *hates* loops
 - May be less of an issue with most recent versions
 - Ideally use matrix operations whenever possible
- Examples:

```
>> mod(A, 3) == 0
```

```
ans =
```

```
0     0     1
0     0     1
```

```
>> A(mod(A,3)==0)
```

```
ans =
```

```
3
6
```

MATLAB tips

A =			
	1	2	3
	4	5	6

- Conventional wisdom: MATLAB *hates* loops
 - May be less of an issue with most recent versions
 - Ideally use matrix operations whenever possible
- Examples:

```
>> A(sum(A,2)>10,2:3)
```

```
ans =
```

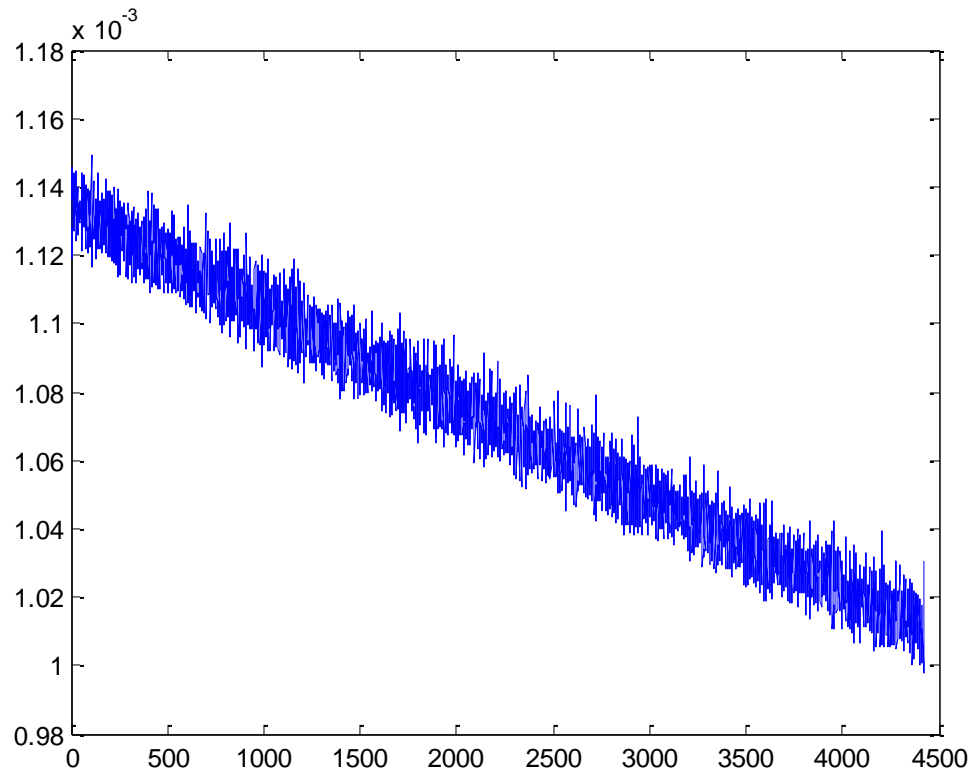
```
5    6
```

Logistic regression implementation

- In my code that solves for w , the only loop is the one that iterates until the weight vector has converged
- My code converges in $\sim 40k$ iterations (35.6s) when
 - $\lambda = 0$
 - w and v initialized to **1**
 - $k = 0.5$
- Can use 'tic' and 'toc' to time your code

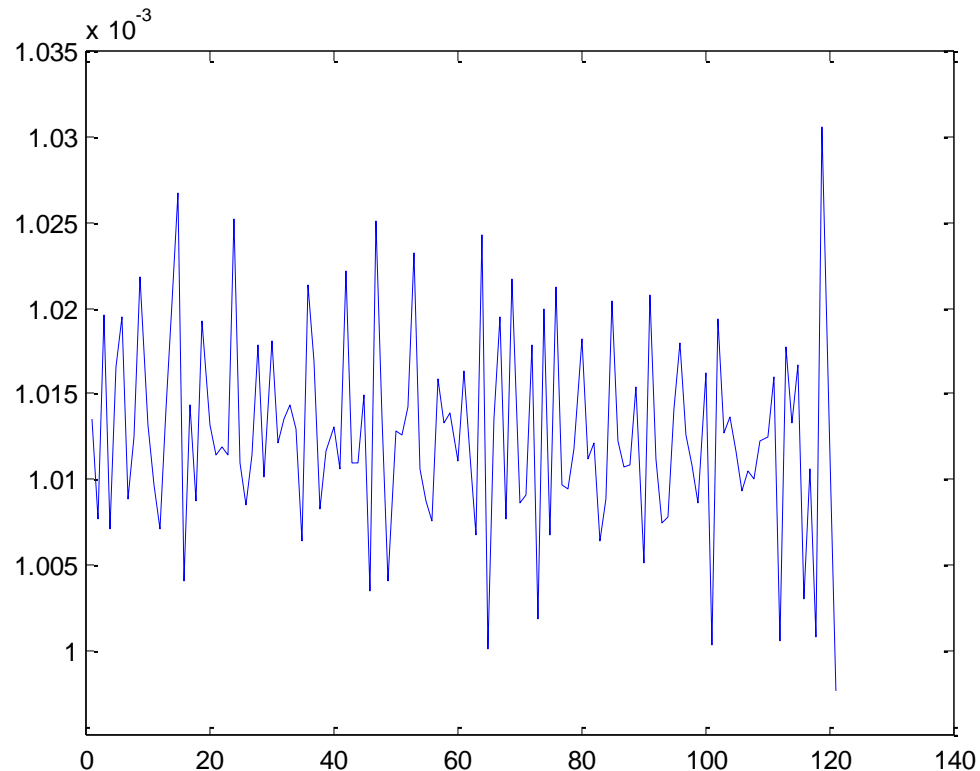
Logistic regression implementation

- Tracking the likelihood and norm of the difference in weight vectors at every iteration can be informative
- L1 regularization, $\lambda = 5$ `>> plot(norms(40000:44420))`



Logistic regression implementation

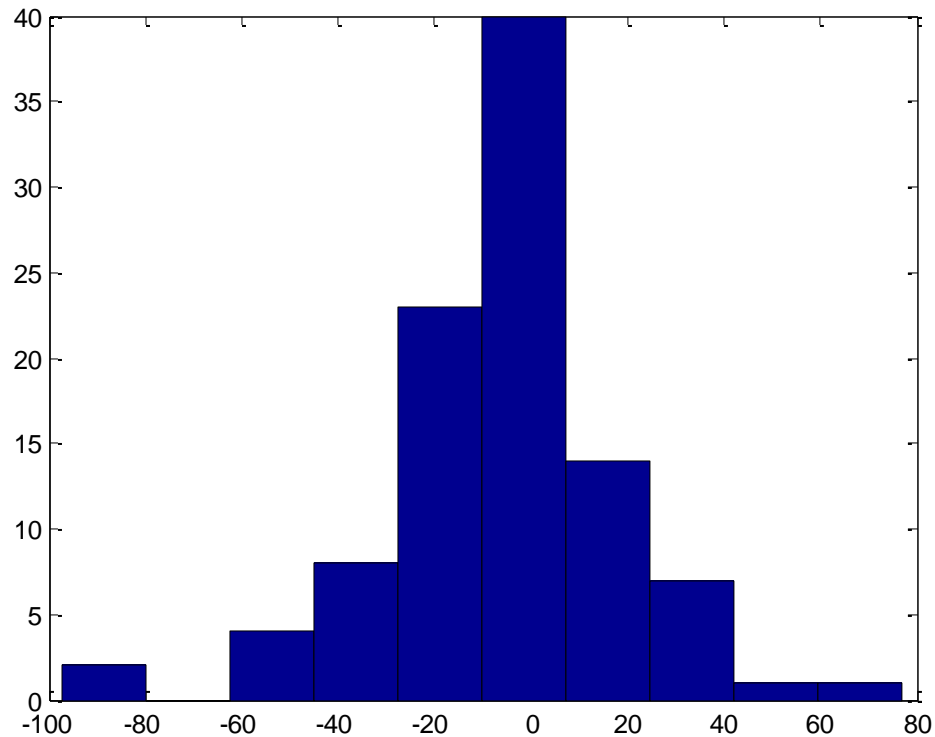
- Tracking the likelihood and norm of the difference in weight vectors at every iteration can be informative
- L1 regularization, $\lambda = 5$ `>> plot(norms(44300:44420))`



Logistic regression implementation

- To test your code, observe what happens to the weight vector as you increase lambda

```
>> hist(w)
```



Bayesian linear regression

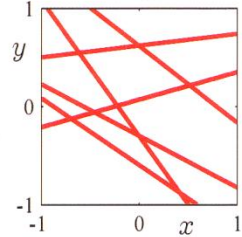
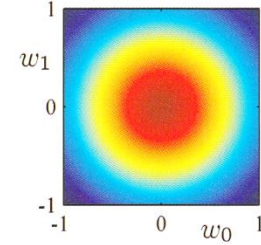
$$y = w_0 + w_1 x$$

No observations

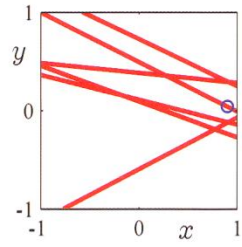
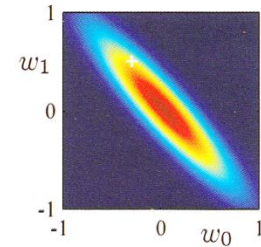
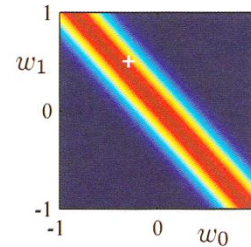
likelihood

prior/posterior

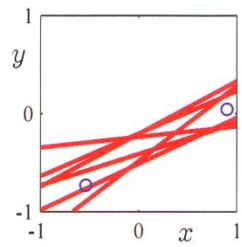
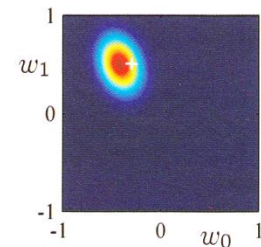
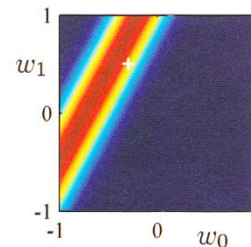
data space



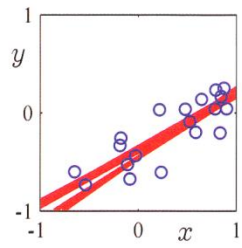
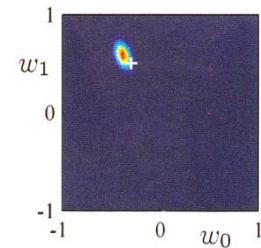
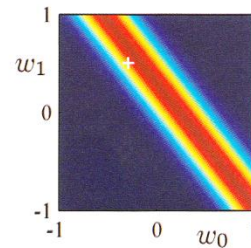
1 observation



2 observations



20 observations



We also discussed...

- Why the Gaussian prior on w in Bayesian linear regression is a conjugate prior, as well as how to compute the posterior
- How to compute information gain in a decision tree with continuous attributes