A Quick Introduction to Machine Learning (Scaling)

Lecturer: John Guttag

6.00.2x

An Example

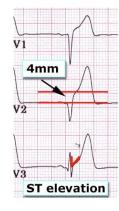
1000 patients with 4 features each

Heart rate in beats per minute Number of past heart attacks ST segment elevation (binary)

Age







Binary outcome based on features

Probabilistic, not deterministic Roughly 31% positive

A Sampling of Examples

```
62.]:1
P0755: [ 48. 1.
                 0.
                     99.1:1
P0383:Γ 103. 1.
P0849: [ 42.
             1.
                 1.
                     92.]:1
P0188:Γ 71.
             2.
                 0. 58.1:0
P0061:Γ 87.
                     79.1:0
                 0.
P0196:Γ 52.
                 0.
                     85.1:0
             0.
P0280: [ 78.
             0.
                 0.
                     81.7:0
P0178: [ 50.
                     59.1:1
                 0.
P0497: [ 80.
                 0. 58.1:0
             0.
P0742:Γ 78.
             2.
                 0.72.1:1
P0527:Γ 78. 1.
                     60.1:0
                 0.
P0915:Γ 60.
                 0.
                     57.1:1
```

Fraction of positives in total population of 1000 was 0.312

Cluster Using K-means (k = 3)

Fraction of positives in population = 0.312

Ran k-means 100 times and chose best clustering

Cluster of size 354 with fraction of positives = 0.338 (1.08x)

Cluster of size 322 with fraction of positives = 0.315 (1.01x)

Cluster of size 324 with fraction of positives = 0.281 (0.90x)

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What Happened?

Features have very different means and variance

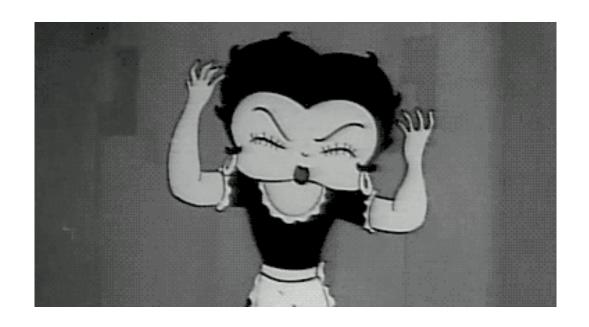
Heart rate in beats per minute (μ = 70, σ = 15) Number of past heart attacks (μ = 0.25, σ = 1.0) ST segment elevation (binary) Age (μ = 65, σ = 15)

HR and age have higher means & greater dynamic range

Euclidean distance will be biased towards them

Does this seem like a good idea?

"No, No, a Thousand Times No"



Rescale Features

Each variable has same mean and variance

$$\chi' = \frac{\chi - \mu_{\chi}}{\sigma_{\chi}} \quad \text{def scaleFeatures(vals):} \\ \text{vals = pylab.array(vals)} \\ \text{mean = sum(vals)/float(len(vals))} \\ \text{sd = stdDev(vals)} \\ \text{vals = vals - mean} \\ \text{return vals/sd}$$

What is the new mean?
What is the new standard deviation?

Testing Scaling

```
def testScaling(n, mean, std):
    vals = []
    for i in range(n):
        vals.append(int(random.gauss(mean, std)))
    print 'original values', vals
    sVals = scaleAttrs(vals)
    print '\n', 'scaled values', sVals
    print '\n', 'new mean =', sum(sVals/len(vals))
    print '\n', 'new sd =', stdDev(sVals)
testScaling(10, 25, 3)
```

Testing Scaling

```
original values [22, 24, 21, 26, 18, 26, 22, 26, 27, 21]
```

```
scaled values [-0.46517657 0.25047969 -0.82300471 0.96613596 -1.89648911 0.96613596 -0.46517657 0.96613596 1.32396409 -0.82300471]
```

new mean = -2.22044604925e-16

new sd = 1.0

The Real Test

Fraction of positives = 0.312

Clustering with unscaled features

Cluster of size 354 with fraction of positives = 0.338 (1.08x)

Cluster of size 322 with fraction of positives = 0.315 (1.01x)

Cluster of size 324 with fraction of positives = 0.281 (0.90x)

Clustering with scaled features

Cluster of size 324 with fraction of positives = 0.055 (0.18x)

Cluster of size 108 with fraction of positives = 0.335 (1.07x)

Cluster of size 568 with fraction of positives = 0.454 (1.45x)