## lazyIris examples

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#### Overview

lazyIris is a small implementation of k-nearest neighbours applied to the famous iris dataset.

## Installing

First, ensure that the devtools package is installed and then install directly from the package github repository.

```
# check for and install devtools.
if(!("devtools" %in% rownames(installed.packages())))
  install.packages("devtools")

# install and load.
if(!("lazyIris" %in% rownames(installed.packages())))
  devtools::install_github("phil8192/lazy-iris")
require(lazyIris)
```

## Loading data

The package has preprocessed iris data attached.

```
attach(iris.data)
```

```
## The following objects are masked from iris.data (pos = 3):
##
## petal.length, petal.width, sepal.length, sepal.width, species
```

Example data may be loaded from the inst/extdata directory by using the *loadData* function. In addition, the *checkData* function will perform any necessary data sanity checks.

```
iris.data <- checkData(loadData())</pre>
## Warning in checkData(loadData()): removed duplicated rows.
##
     cleaned data...
##
     sepal.length
                     sepal.width
                                      petal.length
                                                     petal.width
                                     Min.
##
    Min.
           :4.300
                   Min.
                            :2.000
                                            :1.00
                                                    Min.
                                                            :0.100
   1st Qu.:5.100
##
                    1st Qu.:2.800
                                     1st Qu.:1.60
                                                     1st Qu.:0.300
##
  Median :5.800
                    Median :3.000
                                     Median:4.40
                                                    Median :1.300
##
    Mean
           :5.856
                    Mean
                            :3.056
                                     Mean
                                            :3.78
                                                    Mean
                                                            :1.209
   3rd Qu.:6.400
                                     3rd Qu.:5.10
##
                    3rd Qu.:3.300
                                                     3rd Qu.:1.800
##
   Max.
           :7.900
                    Max.
                            :4.400
                                     Max.
                                            :6.90
                                                     Max.
                                                            :2.500
##
               species
##
   Iris-setosa
                   :48
##
   Iris-versicolor:50
    Iris-virginica:49
##
##
```

### A quick look at the data

##

The dataset consists of 4 features and 3 possible classes. Some of the features are highly correlated:

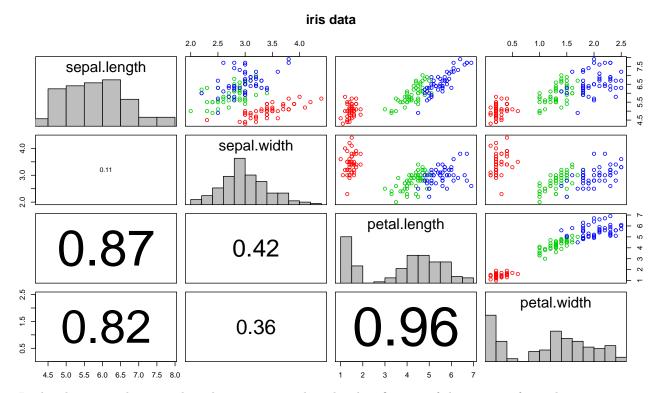
```
cor(iris.data[, 1:4])
```

	sepal.length	sepal.width	petal.length	petal.width
sepal.length	1.0000000	-0.1093208	0.8713046	0.8170583
sepal.width	-0.1093208	1.0000000	-0.4210574	-0.3563762
petal.length	0.8713046	-0.4210574	1.0000000	0.9618828
petal.width	0.8170583	-0.3563762	0.9618828	1.0000000

#### Iris data visualisation

The package provides a means to visualise the relationship between the 4 features and the corresponding class.

```
# plot all the data.
visualise(iris.data, class.name="species", main="iris data", plot.hist=TRUE,
    plot.cor=TRUE)
```



In the above visualisation, the colours correspond to the classification of the species of iris plant:

colour	species
$\operatorname{red}$	Iris setosa
green	Iris versicolour
blue	Iris virginica

The lower left panels show the correlation between the 4 iris features, the diagonal panels contain a histogram of the distribution of each feature, and finally, the upper right panels contain scatter plots of each possible feature combination colour coded by species.

## Querying the data

The knn function makes it possible to query the data for neighbouring instances given an arbitrary list of features.

#### 10-Nearest neighbours

The following example obtains the *top 10* nearest neighbours to query:

```
# form the query.
# in this example, the feature values are actually the mean values in the
# dataset, thus the results may be interpreted as the top 10 "most average"
# instances.

query <- list(
    sepal.length=5.84,
    sepal.width=3.05,
    petal.length=3.76,
    petal.width=1.20)

# obtain the nearest-neighbours.
top.10 <- knn(query, iris.data, 10)
print(top.10, row.names=FALSE)</pre>
```

sepal.length	sepal.width	petal.length	petal.width	species	distance
5.6	2.9	3.6	1.3	Iris-versicolor	0.3401470
5.8	2.7	3.9	1.2	Iris-versicolor	0.3790778
5.6	3.0	4.1	1.3	Iris-versicolor	0.4309292
6.1	2.8	4.0	1.3	Iris-versicolor	0.4446347
5.7	2.8	4.1	1.3	Iris-versicolor	0.4557412
5.7	3.0	4.2	1.2	Iris-versicolor	0.4644351
5.7	2.9	4.2	1.3	Iris-versicolor	0.4956813
5.8	2.6	4.0	1.2	Iris-versicolor	0.5115662
5.8	2.7	4.1	1.0	Iris-versicolor	0.5288667
5.9	3.0	4.2	1.5	Iris-versicolor	0.5382379

#### Classification

In addition to the N-nearest neighburs, the function also returns the distance from the query point. This distance can be used to predict the most likely class of the query point using the *classifier* function.

```
## [1] "prediction = Iris-versicolor confidence = 1"
```

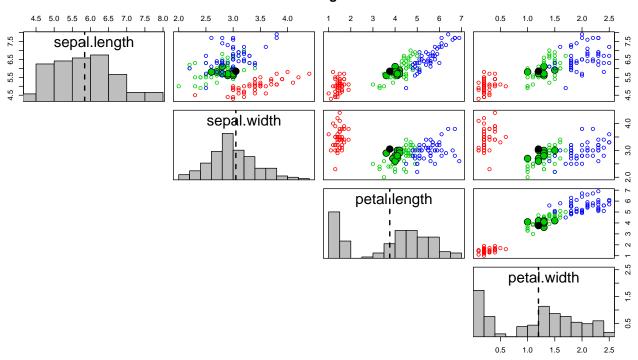
#### Visualising a query result

Given a list of nearest neighbours returned from the knn function, it is possible to visualise the query point and it's nearest neighbours over all dimensions in the feature space by using the visualise function.

```
top.10 <- knn(query, iris.data, k=10)

# visualise the result.
visualise(iris.data, class.name="species", query=q, neighbours=top.10,
    main="iris data neighbours", plot.hist=TRUE, plot.cor=FALSE)</pre>
```

#### iris data neighbours



In the above plot, the query point is shown as a black point. The resulting neighbours from the knn query are highlighted (opaque) circles. In addition, the query point with respect to the distribution of features has been highlighted with a black dashed vertical line over the corresponding feature histograms.