

# Statistical Modeling and Machine Learning

## 01 - Math refresher

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### 1 Reading

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#### 1.1 Maths

You can find a concise summary of the required math background in appendix A.1-A.2 of the freely available book by Barber: <http://web4.cs.ucl.ac.uk/staff/D.Barber/pmwiki/pmwiki.php?n=Brml.Online>. (Appendix A.3 to A.6 can also become helpful later.). Alternatively, you can use Chapter 2 up to section 2.11 determinant included from [http://deeplearningbook.org/contents/linear\\_algebra.html](http://deeplearningbook.org/contents/linear_algebra.html)

Online lectures, exercises and exam questions for linear algebra can be found here: <http://ocw.mit.edu/courses/mathematics/18-06-linear-algebra-spring-2010/>

Relevant videos are also available at the coursera machine learning lecture (Section III. Linear Algebra Review): <https://class.coursera.org/ml-005/lecture>

#### 1.2 R programming

Get familiar with R basics by going through the (very) short introduction to R by Paul Torfs and Claudia Brauer. <http://cran.r-project.org/doc/contrib/Torfs+Brauer-Short-R-Intro.pdf>. The TODO sections in this tutorial are recommended but you do not need to hand in the solutions.

### 2 Calculating with matrices and gradients

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**Question 1:** Let  $\mathbf{A} \in \mathbb{R}^{m \times n}$  and  $\mathbf{B} \in \mathbb{R}^{n \times l}$  be matrices. Show by writing componentwise that

$$(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T \quad (1)$$

**Question 2:** Show for invertible matrices  $\mathbf{A}$  and  $\mathbf{B}$  that

$$(\mathbf{AB})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1} \quad (2)$$

by using the definition of  $(\mathbf{AB})^{-1}$  as the matrix that gives the identity matrix when multiplied from left and right with  $\mathbf{AB}$ .

In the following, we will use an alternative notation for the gradient:

$$\nabla_{\mathbf{x}} f(\mathbf{x}) = \frac{\partial f(\mathbf{x})}{\partial \mathbf{x}}. \quad (3)$$

**Question 3:** Let  $\mathbf{x}$  and  $\mathbf{b}$  be vectors,  $\mathbf{A}$  a matrix. Show componentwise that

$$\frac{\partial (\mathbf{b}^T \mathbf{x})}{\partial \mathbf{x}} = \mathbf{b} \quad (4)$$

**Question 4:** Let  $\mathbf{A}$  be a symmetric matrix. Show by componentwise partial derivatives that

$$\frac{\partial}{\partial \mathbf{x}} \left( \frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x} \right) = \mathbf{A} \mathbf{x} \quad (5)$$

**Question 5:** Let  $\mathbf{x}$  be a vector in  $R^D$ ,  $\mathbf{A}$  a matrix in  $R^{D \times D}$ . Show the following equations by writing them componentwise:

Show that

$$\nabla_{\mathbf{x}} \|\mathbf{x}\| = \mathbf{x} / \|\mathbf{x}\| \quad (6)$$

**Question 6:** Show that

$$\mathbf{P} = \frac{\mathbf{a} \mathbf{a}^T}{\|\mathbf{a}\|^2} \quad (7)$$

is the projection operator onto the one-dimensional vector space spanned by  $\mathbf{a}$ . Hint: Any vector  $\mathbf{x} \in R^D$  can be split into two parts: the part within  $\text{span}(\mathbf{a})$ ,  $\alpha \mathbf{a}$ , and the part orthogonal to it,  $\mathbf{x}_{\perp}$ .

### 3 Optimization

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**Question 7:** In R, plot the function  $f(x) = x \exp(-\frac{(x-1)^2}{2})$  for 100 points in  $[-2, 2]$  using `x=seq(-2, 2, length.out=100)`. How many local extrema do you see? Compute analytically their values and mark the solution by vertical lines using the function `abline(v=...)`.

## 4 Data exploration in R

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The famous (Fisher's or Anderson's) iris data set ([https://en.wikipedia.org/wiki/Iris\\_flower\\_data\\_set](https://en.wikipedia.org/wiki/Iris_flower_data_set)) gives the measurements in centimeters of the variables sepal length and width and petal length and width, respectively, for 50 flowers from each of 3 species of iris. The species are Iris setosa, versicolor, and virginica. The data is stored as a data.frame, which is a matrix like object. The main advantage is the possibility to store multiple different types together, whereas a matrix can only store one type (e.g integer). This is possible, because the object is internally stored as a list of elements of equal length. Some essential functions to explore and view the data are:

`str()`: Gives the structure of an object

`head()/tail()`: to view the first/last six rows of the dataset

`myData$myColumn`: Access the column of a data.frame

`myData[rowvector, columnvector]`: access specific rows and columns of the data.frame by number

`myData[rowname, columnname]`: access specific rows and columns of the data.frame by name

`myData$newColumn = value`: add a new column (must be of same length as the other)

`?myObject`: Get help for an object

```
data("iris")
str(iris)

## 'data.frame': 150 obs. of  5 variables:
##  $ Sepal.Length: num  5.1 4.9 4.7 4.6 5 5.4 4.6 5 4.4 4.9 ...
##  $ Sepal.Width : num  3.5 3 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1 ...
##  $ Petal.Length: num  1.4 1.4 1.3 1.5 1.4 1.7 1.4 1.5 1.4 1.5 ...
##  $ Petal.Width : num  0.2 0.2 0.2 0.2 0.2 0.4 0.3 0.2 0.2 0.1 ...
##  $ Species      : Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 1 ...

head(iris)

##   Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 1         5.1         3.5         1.4         0.2   setosa
## 2         4.9         3.0         1.4         0.2   setosa
## 3         4.7         3.2         1.3         0.2   setosa
## 4         4.6         3.1         1.5         0.2   setosa
## 5         5.0         3.6         1.4         0.2   setosa
## 6         5.4         3.9         1.7         0.4   setosa

tail(iris)

##   Sepal.Length Sepal.Width Petal.Length Petal.Width Species
## 145         6.7         3.3         5.7         2.5 virginica
## 146         6.7         3.0         5.2         2.3 virginica
## 147         6.3         2.5         5.0         1.9 virginica
## 148         6.5         3.0         5.2         2.0 virginica
```

```
## 149      6.2      3.4      5.4      2.3 virginica
## 150      5.9      3.0      5.1      1.8 virginica

head(iris$Species)

## [1] setosa setosa setosa setosa setosa setosa
## Levels: setosa versicolor virginica

iris[1:10,2]

## [1] 3.5 3.0 3.2 3.1 3.6 3.9 3.4 3.4 2.9 3.1

head(iris[, "Sepal.Width"])

## [1] 3.5 3.0 3.2 3.1 3.6 3.9
```

Note that, the column "Species" is of class Factor, which appears like a string/character object but differs significantly, as it internally stores all levels (in this case three) as integers and associates each level with a label.

```
chrObject = as.character(iris$Species)
str(iris$Species)

## Factor w/ 3 levels "setosa","versicolor",...: 1 1 1 1 1 1 1 1 1 1 ...

str(chrObject)

## chr [1:150] "setosa" "setosa" "setosa" "setosa" "setosa" ...

levels(iris$Species)

## [1] "setosa"      "versicolor" "virginica"

str(as.numeric(iris$Species))

## num [1:150] 1 1 1 1 1 1 1 1 1 1 ...

str(as.numeric(chrObject))

## Warning in str(as.numeric(chrObject)): NAs introduced by coercion

## num [1:150] NA NA NA NA NA NA NA NA NA NA ...
```

**Question 8:** Given the "iris" dataset, answer the following questions: For each variable (except Species), what is the mean, the median, the interquartile range (range of the middle 50% the data) and the entire range of the data?

Hint: Five-number summary

**Question 9:** Plot a histogram of Sepal.Length with the default breaks and with 20 breaks.

Hint: ?hist

**Question 10:** How many of each of the flower species have a sepal length smaller than 5.5 and how many more?

Hint: New groups can be easily attached as new columns and `?table`

**Question 11:** Plot Sepal.Length against Petal.Length, the finding from 2.), every variable against each other and all the numeric variables against each other.

Hint: `?plot` and `?pairs`

**Question 12:** Produce separate plots of Petal.Width vs. Petal.Length for each Species combined in one plotting window.

Hint: `?par` and `mfrow` argument

## 5 Projections in 3D

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**HOMEWORK (This is a homework exercise only and will not be covered in class. It is still due to submission.)**

Define a pyramid as a scaffold of points along its edges:

```
p1= c(-1,-1,0)
p2= c(-1,1,0)
p3= c(1,1,0)
p4= c(1,-1,0)
p5= c(0,0,2)

v = seq(0,1,0.05)
e1 = v %>% (p2-p1) + rep(1,21) %>% p1
e2 = v %>% (p3-p2) + rep(1,21) %>% p2
e3 = v %>% (p4-p3) + rep(1,21) %>% p3
e4 = v %>% (p1-p4) + rep(1,21) %>% p4
e5 = v %>% (p5-p1) + rep(1,21) %>% p1
e6 = v %>% (p5-p2) + rep(1,21) %>% p2
e7 = v %>% (p5-p3) + rep(1,21) %>% p3
e8 = v %>% (p5-p4) + rep(1,21) %>% p4

pyramid = rbind(e1,e2,e3,e4,e5,e6,e7,e8)
```

**Question 13:** Write down the matrix that projects from  $(x, y, z)$  to  $(x, y)$  and plot the  $(x, y)$  points of the pyramid. Be careful about the order and dimensions of the matrices that you multiply.

**Question 14:** Now define two rotation matrices in two dimensions, R1 and R2, one rotating around the z axis by phi and the other rotating around the y axis by theta. Use them to rotate the pyramid and to project in onto the  $x$ - $y$  plane. Try different angles phi and theta, e.g.  $\text{phi} = 0.05 \cdot \pi$ ,  $\text{theta} = 0.3 \cdot \pi$ .

## 6 Reading

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### 6.1 Assignment

Read Bishop's book up to section 1.2.4 (The Gaussian distribution) included. Optionnally, you can read sections 1.2.5-6.

### 6.2 Remarks

We will not look into reinforcement techniques.

We will look into the bootstrap later.

Figure 1.8 and Table 1.2 are not consistent, but the argumentation is right.

Equation (1.27): Note that  $g$  has to be a bijection to be a change of variable.

Equation (1.33): Note that  $f$  can be vector-valued.

Equation (1.37): We prefer the notation  $E_{X|Y}(f(X)) = \dots$

### 6.3 Questions

**Question 15: (At home only)** What is supervised learning and what is unsupervised learning? Give some examples.