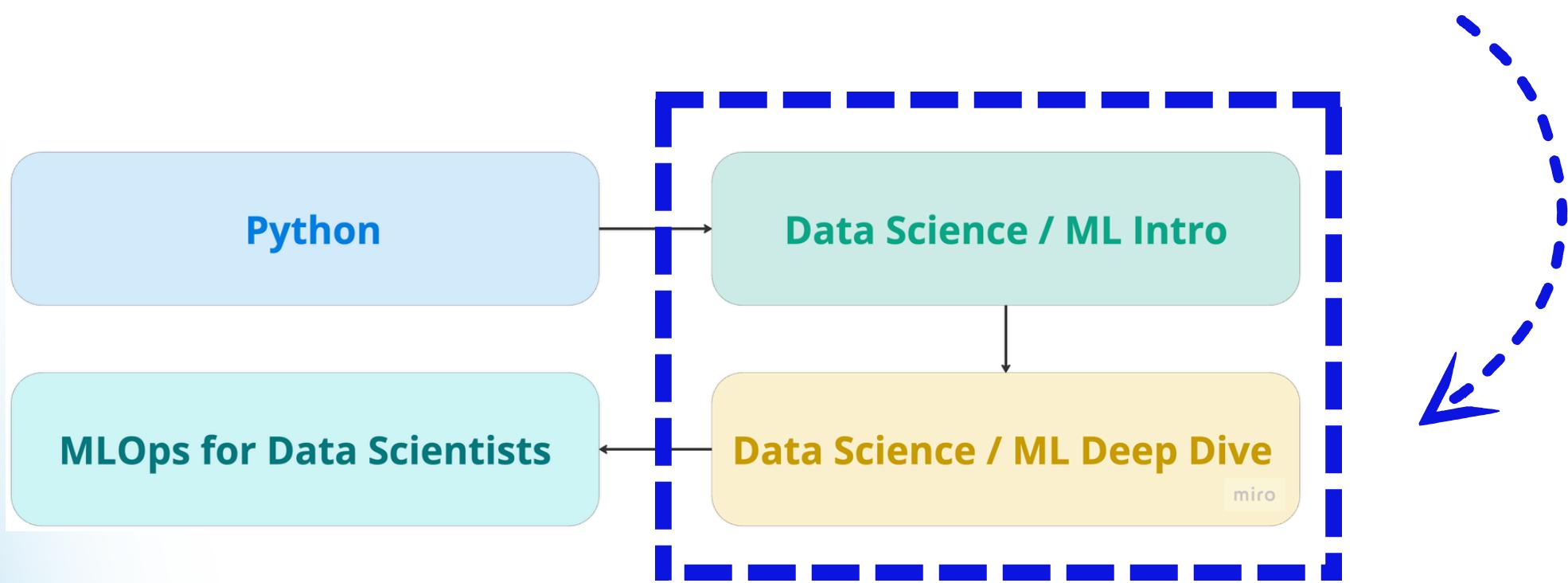




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# ML ROADMAP with **LINKS** from **JUNIOR TO SENIOR**



([Link to the full roadmap](#) at the end!)

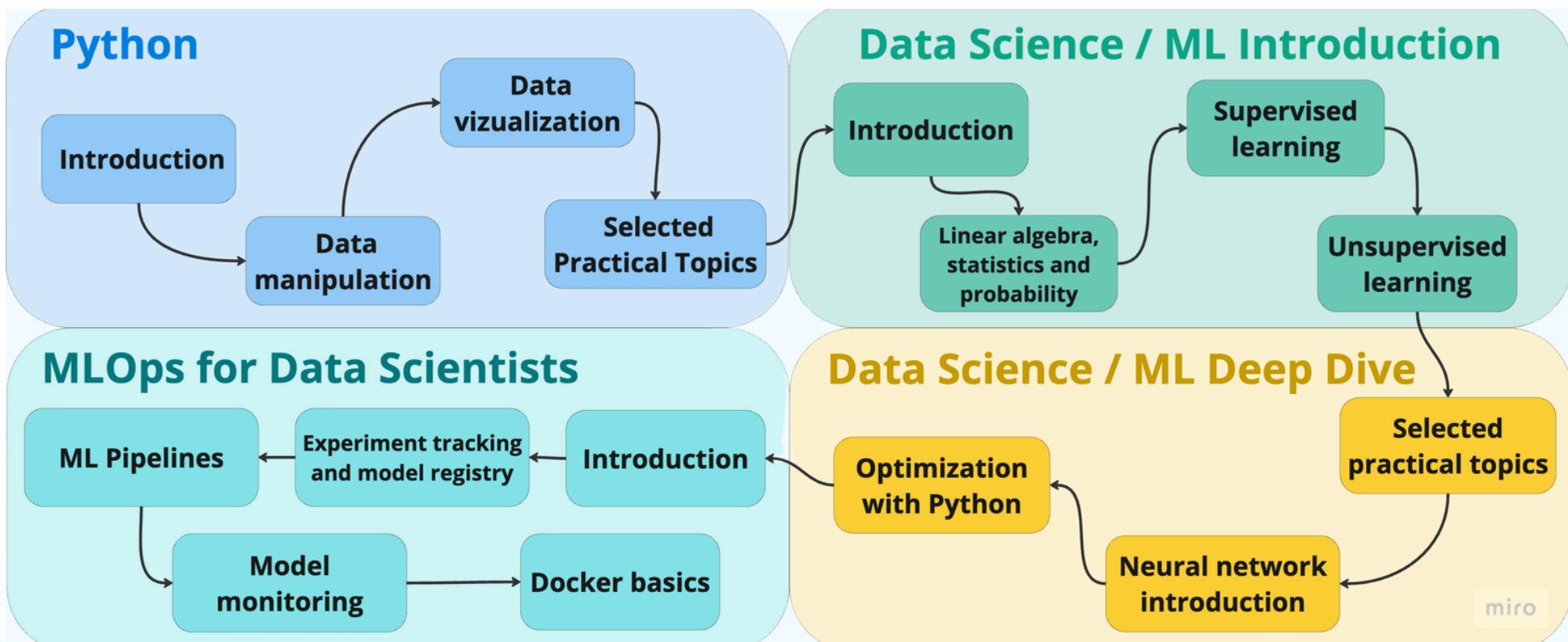


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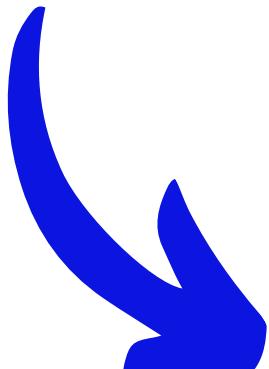
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The entire roadmap can be  
downloaded and  
completed for  
**ABSOLUTE FREE.**



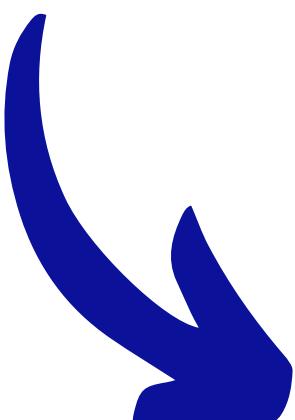
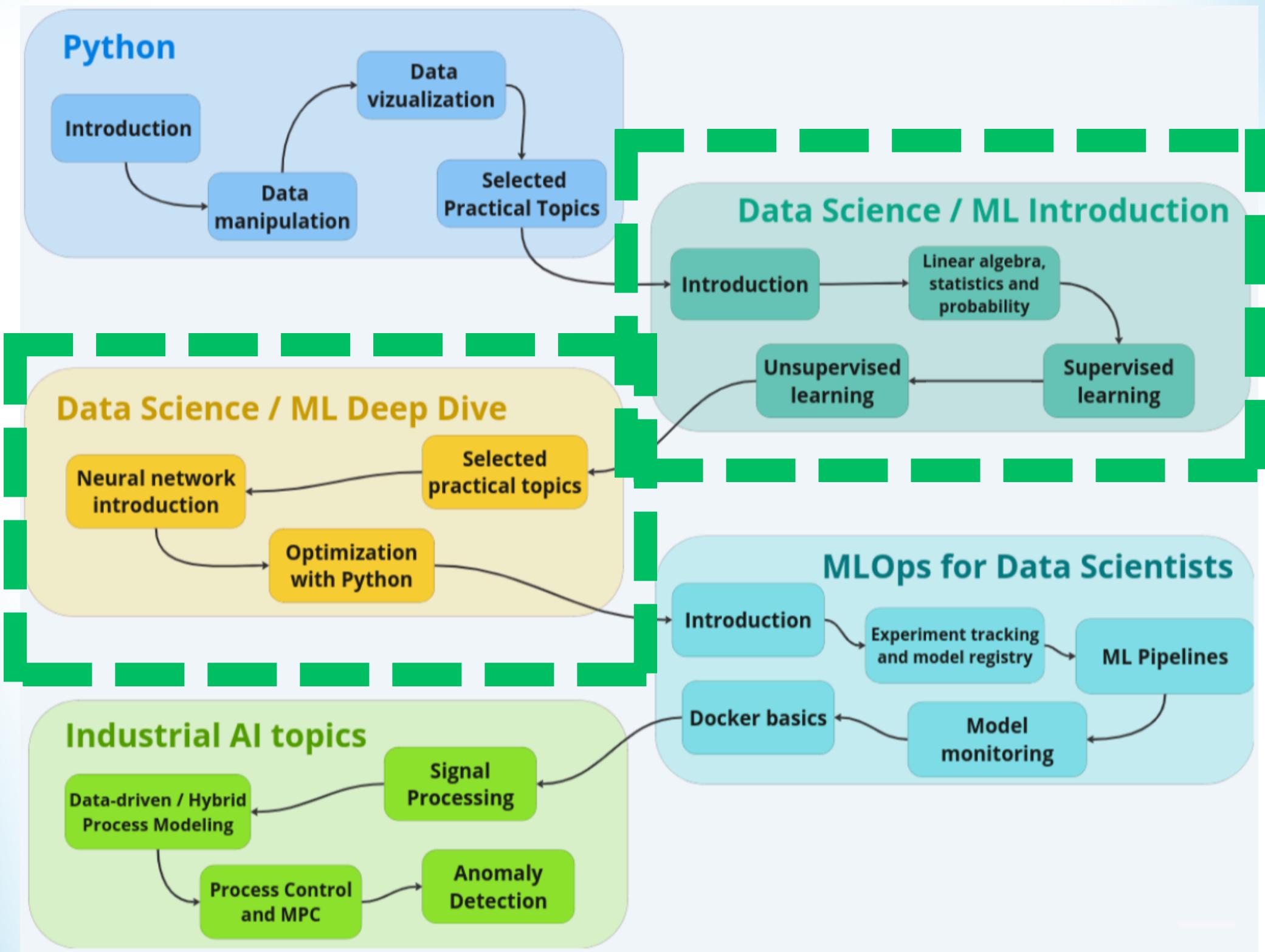
[LINK TO THE DS ROADMAP](#)

Ok, let's check the breakdown!





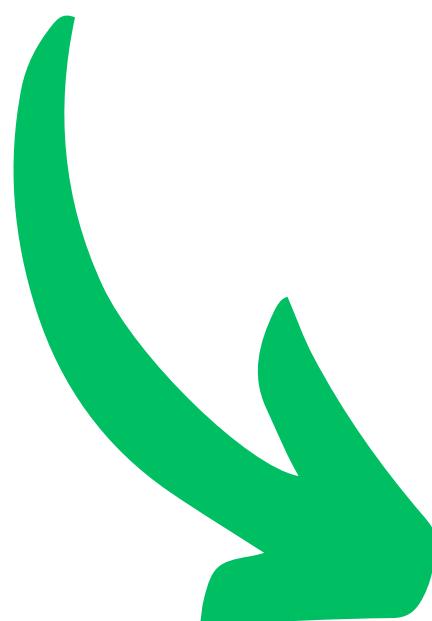
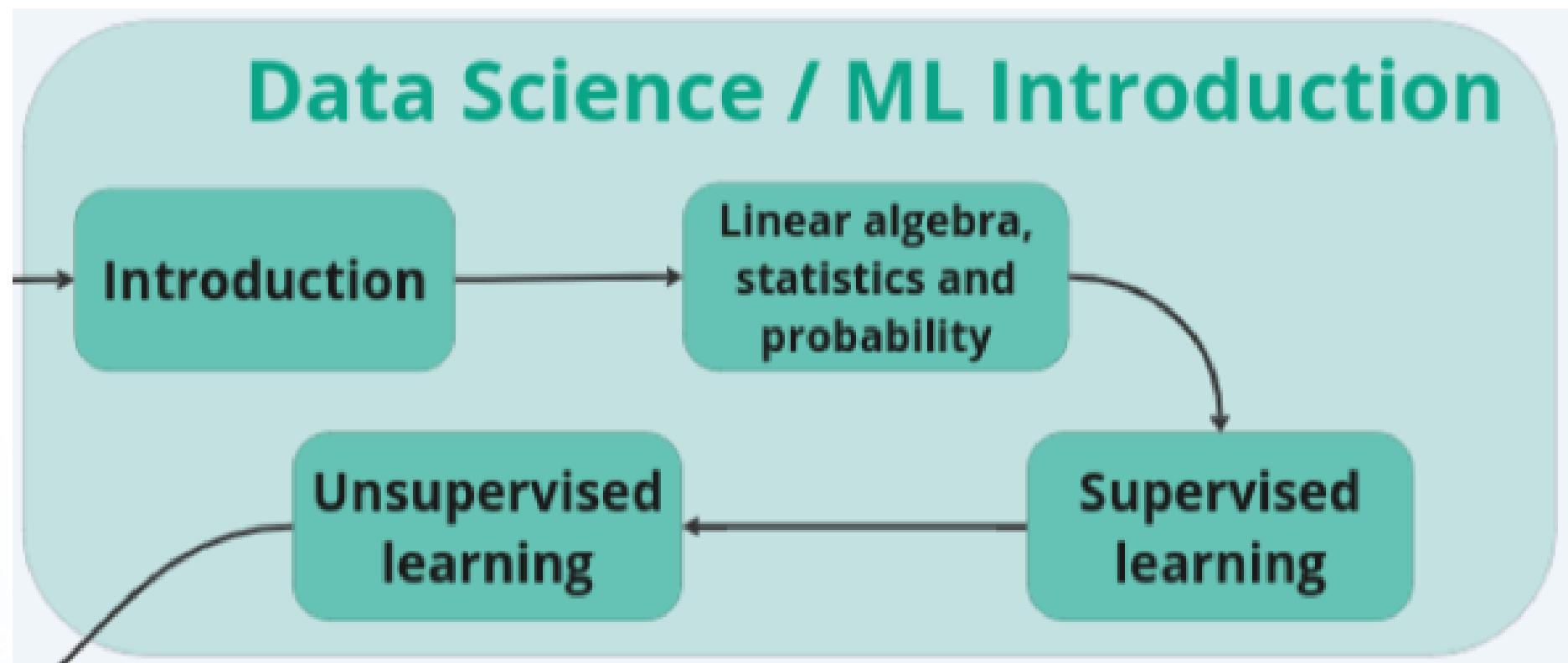
# From the full roadmap, here we cover the ML part.





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# Let's start from Introduction!





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1

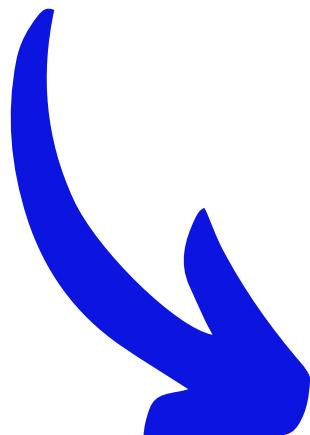
# Introduction

**Andrew Ng course:** [Link](#)

This is a perfect course to get an overview of what machine learning is and what are the two most common problems that are solved by ML: regression and classification.

1.9

**See next...**





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2

## Probability, statistics, and linear algebra

### Linear Algebra

**Step 1**

Videos of 3blue1brown about linear algebra: [Link](#)

**Step 2**

Tutorial of Python Linear Algebra by Pablo Caceres: [Link](#)

### Probability and Statistics

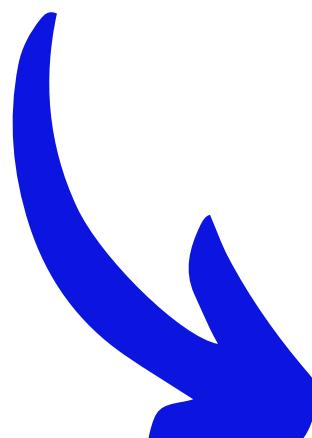
**Step 1**

Statistics Crash Course by Adriene Hill: [Link](#)

**Step 2**

Learn Statistics with Python by Ethan Weed: [Link](#)

See next...





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3

## Supervised learning

# Linear regression

### Step 1. Intro theory

Nando de Freitas lectures at UBC

- [Lecture 1](#)
- [Lecture 2](#)

### Step 2. Python Implementation:

- [Closed-form linear regression implementation](#)
- [Linear regression with gradient descent](#)

### Step 3. Regularization in linear regression

1: Nando de Freitas lectures at UBC

- [Lecture 1](#)
- [Lecture 2](#)

2: Visual explanation with code

- [Explanation 1](#)
- [Explanation 2](#)

### Step 4. Sklearn tutorial with Lasso model

- [LINK](#)

See next...





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## Supervised learning

### Logistic regression

#### Step 1. Intro

Logistic regression topic of mlcourse.ai

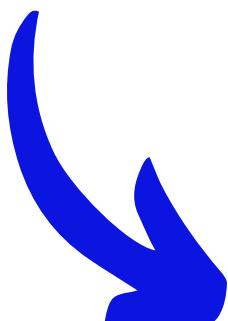
- [LINK](#)

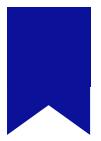
#### Step 2. Odds ratio as weights interpretability

- [LINK](#)

1.8

See next...





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## Supervised learning

### Gradient Boosting

#### Step 1. Intro

Gradient Boosting topic of [mlcourse.ai](#)

- [LINK](#)

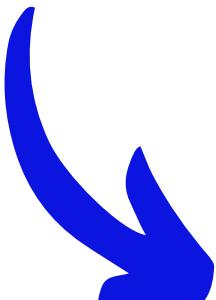
#### Step 2. Deeper dive

- [Tutorial by Alexey Natekin](#)
- [XGBoost original paper](#)

#### Step 3. Demo playground by Alex Rogozhnikov

- [Demo 1](#)
- [Demo 2](#)

See next...





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## Supervised learning

### Random Forest

#### Step 1. Intro

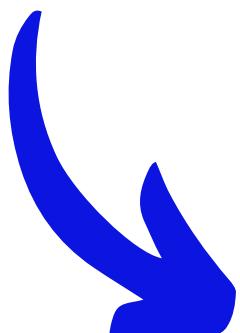
#### Lectures by Nando de Freitas

- [Lecture 1](#)
- [Lecture 2](#)
- [Lecture 3](#)

#### Step 2. Bagging topic on mlcourse.ai

- [LINK](#)

See next...





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## Supervised learning

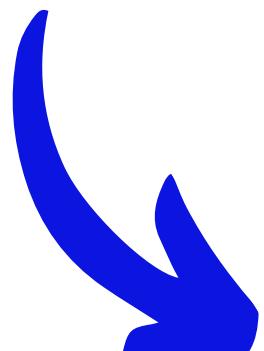
### k Nearest Neighbours (k-NN)

Guide by Mustafa Murat ARAT

- [LINK](#)

1.9

See next...





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4

## Unsupervised learning

### Clustering

#### k-Means

Guide by Mustafa Murat ARAT

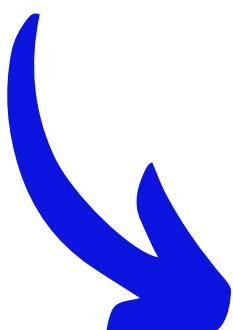
- [LINK](#)

#### DBSCAN

Detailed Applied Guide

- [LINK](#)

See next...





4

## Unsupervised learning

### Dimentionality Reduction

#### PCA

- Material from Sebastian Rashka

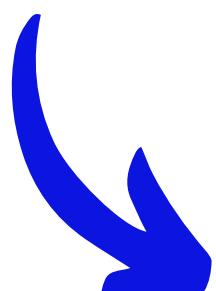
#### t-SNE

- What is it and how to run it in Python
- How to use t-SNE effectively\_(with great visualizations)

#### UMAP

- Detailed guide

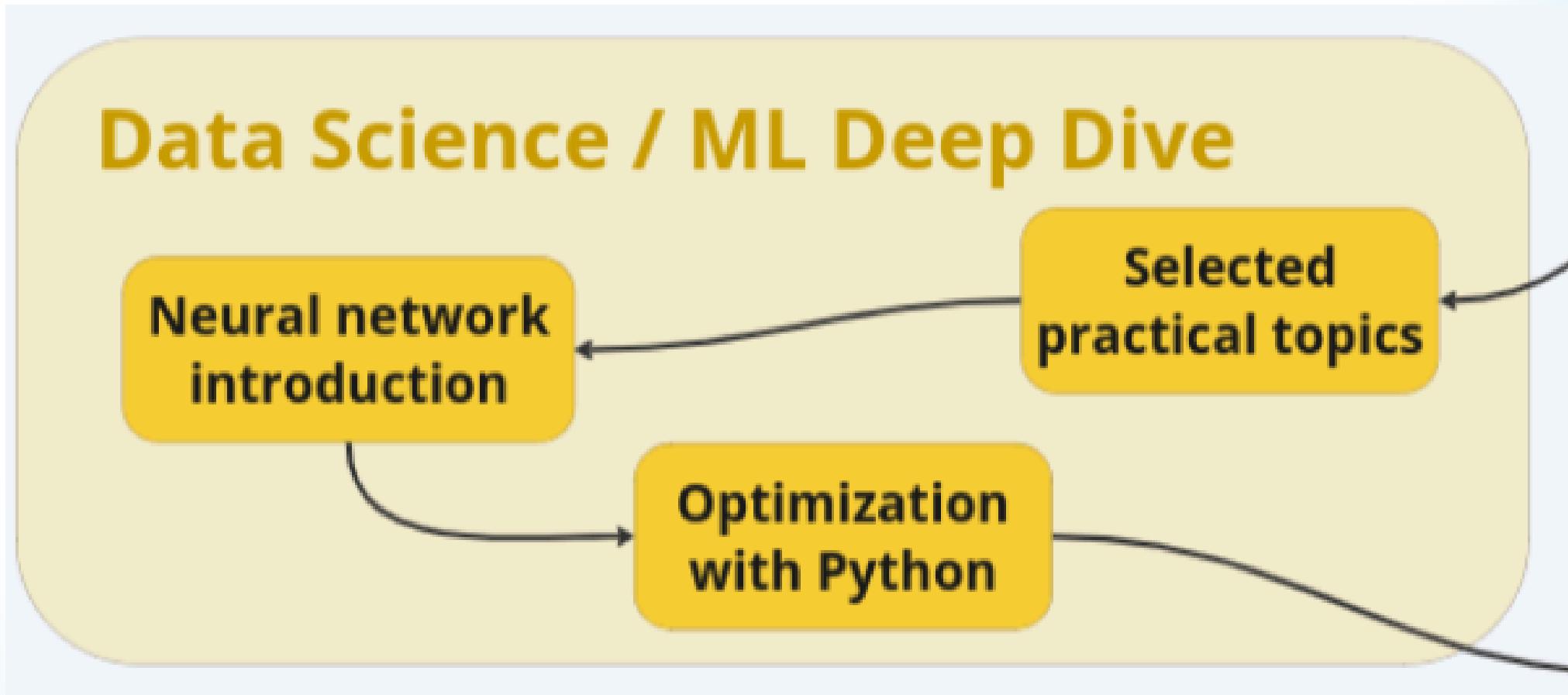
See next...





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# Now, deeper dive...



1.9





1

# Selected Practical Topics

## Feature selection

- [Detailed Guide](#)

## Feature importance

### Linear Methods

- [Guide 1](#)
- [Guide 2](#)

### Tree-based Methods

- [Video Guide](#)

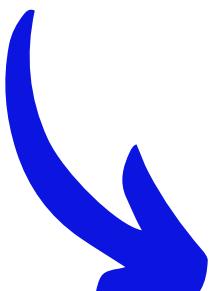
### Permutation feature importance

- [Guide](#)

### SHAP

- [Guide](#)

## See next...





1

# Selected Practical Topics

## Model metrics evaluation

### Regression Metrics

- [Detailed Article](#)

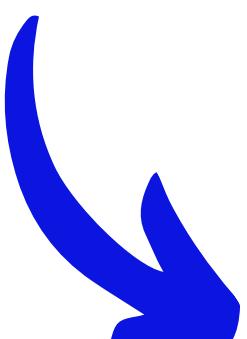
### Classification Metrics

- [Detailed Article](#)

## Cross Validation

- [Detailed Article](#)

See next...





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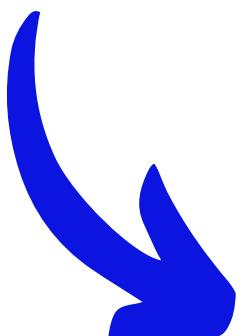
## Neural Networks Introduction

### 5 Courses of Coursera Deep Learning Specialization

- [LINK](#)

To get an intro into the topic, Andrew Ng's specialization is still great. He goes step-by-step and I guarantee you will understand the concept. From that, you can go deeper depending on the domain you are interested in.

See next...





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## Optimization with Python

### Introduction to mathematical optimization with Python

- [Detailed Guide](#)

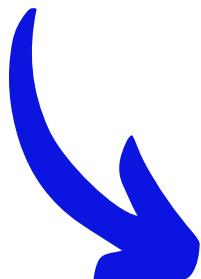
### Bayesian Optimization

- [Playground with theory explanation](#)
- [Deep theory dive by Nando de Freitas](#)

### Optimization with SciPy

- [Link 1](#)
- [Link 2](#)
- [Link 3](#)
- [Link 4](#)

See next...





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## Optimization with Python

### Playground

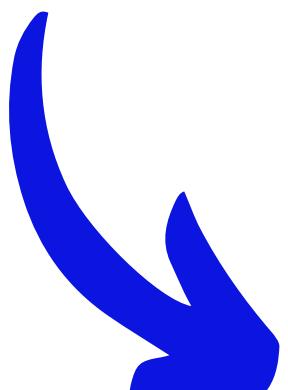
- [Detailed Guide](#)

### Additional Resources

- [Book by Nocedal](#)
- [Extensive list of optimization resources](#)

1.3

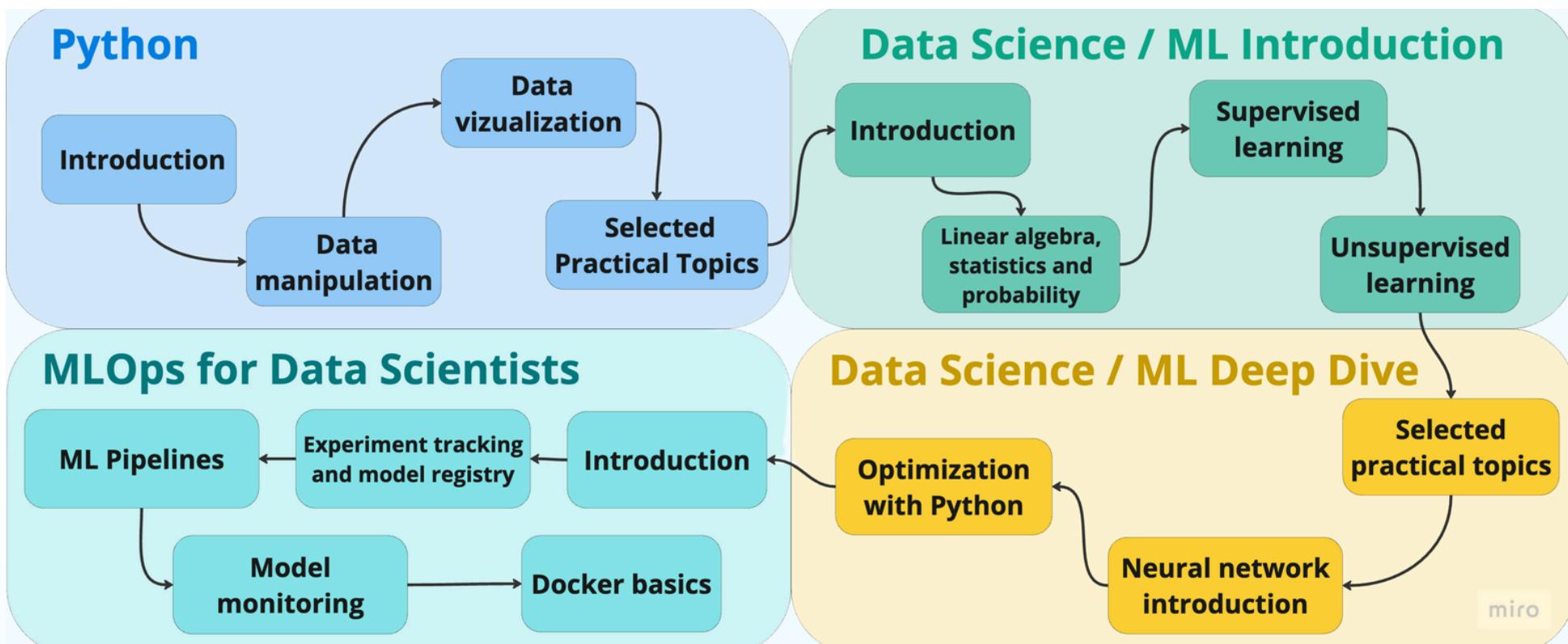
See next...





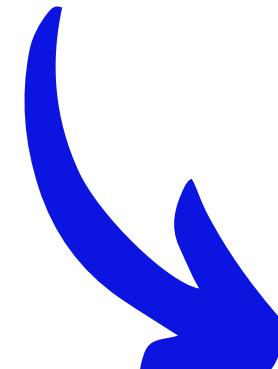
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The entire roadmap can be  
downloaded and  
completed for  
**ABSOLUTE FREE.**



**LINK TO THE DS ROADMAP**

Finally...





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## so I know that the content resonates!

1.8



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