## The Data Engineering Cookbook

How to master the plumbing of data science

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## Contents

1	Intr	oducti	on	(			
2	What is Data Science?						
	2.1	Data S	Scientist				
	2.2	Data I	Engineer				
	2.3	Data A	Analyst				
3	The	Basic	Skills				
4	Lea	rn to V	Write Code				
	4.1	Coding	g Basics				
	4.2	Learn	To Use GitHub				
	4.3	Agile I	Development				
		4.3.1	Scrum				
		4.3.2	OKR				
5	Con	nputer	Science Basics				
	5.1	Learn	how a Computer Works				
		5.1.1	CPU,RAM,GPU,HDD				
		5.1.2	Differences between PCs and Servers				
	5.2	Compi	uter Networking				
		5.2.1	ISO/OSI Model				
		5.2.2	IP Subnetting				
		5.2.3	Switch, Level 3 Switch				
		5.2.4	Router				
		5.2.5	Firewalls				
	5.3	Securit	ty and Privacy				
		5.3.1	SSL Public & Private Key Certificates				
		5.3.2	What is a certificate authority				
		5.3.3	JAva Web Tokens	-			
		5.3.4	GDPR regulations				
		5.3.5	Privacy by design				
	5.4	Linux		]			
		5.4.1	OS Basics	1			
		5.4.2	Shell scripting				
		5.4.3	Cron jobs				
		5.4.4	Packet management				

6	Dat	a Scie	nce Platform
	6.1	Securi	ity Zone Design
		6.1.1	How to secure a multi layered application
		6.1.2	Cluster security with Kerberos
	6.2	Lamb	da Architecture
		6.2.1	Stream and Batch processing
		6.2.2	My Big Data Platform Blueprint
		6.2.3	Three methods of streaming
	6.3	Big D	ata
		6.3.1	What is big data and where is the difference to data science and
			data analytics?
		6.3.2	The 4Vs of Big data:
		6.3.3	Why Big Data?
		6.3.4	What are the tools associated?
	6.4	What	is the difference between a Data Warehouse and a Data Lake
	6.5	Hadoo	op Platfroms
		6.5.1	What makes Hadoop so popular?
		6.5.2	How does a Hadoop System architecture look like
		6.5.3	What tools are usually in a with Hadoop Cluster
		6.5.4	How to select Hadoop Cluster Hardware
	6.6	Is ET	L still relevant for Analytics?
	6.7		Cloud
		6.7.1	AWS, Azure, IBM, Google Cloud basics
		6.7.2	cloud vs on premise
		6.7.3	up & downsides
		6.7.4	Security
	6.8		er
	0.0	6.8.1	What is docker and what do you use it for
		6.8.2	How to create, start, stop a Container
		6.8.3	Docker micro services?
		6.8.4	Kubernetes
		6.8.5	Why and how to do Docker container orchestration
		0.0.0	why and now to do Docker container orchestration
7	Hov	w to In	ngest Data
	7.1	Applie	cation programming interfaces
		7.1.1	REST APIs
		7.1.2	HTTP Post/Get
		7.1.3	API Design
		7.1.4	Implementation

	7.2	7.1.5 JSON	OAuth security	27 27
	1.4	7.2.1	Curan important for DECT ADIa	27
			Super important for REST APIs	
		7.2.2	How is it used for logging and processing logs	27
8	Dist		d Processing	27
	8.1	MapRe	educe	28
		8.1.1	Why was MapReduce Invented	28
		8.1.2	How does that work	28
		8.1.3	What is the limitation of MapReduce?	28
	8.2	Apach	e Spark	28
		8.2.1	Spark Basics	29
		8.2.2	What is the difference to MapReduce?	29
		8.2.3	How to do stream processing	29
		8.2.4	How to do batch processing	29
		8.2.5	How does Spark use data from Hadoop	29
		8.2.6	What is a RDD and what is a DataFrame?	29
		8.2.7	Spark coding with Scala	29
		8.2.8	Spark coding with Python	29
		8.2.9	How and why to use SparkSQL?	29
		8.2.10	Machine Learning on Spark? (Tensor Flow)	29
	8.3	Messag	ge queues with Apache Kafka	29
		8.3.1	Why a message queue tool?	29
		8.3.2	Kakfa architecture	29
		8.3.3	What are topics	29
		8.3.4	What does Zookeeper have to do with Kafka	29
		8.3.5	How to produce and consume messages	29
	8.4	Machin	ne Learning	29
		8.4.1	Training and Applying models	30
		8.4.2	What is deep learning	30
		8.4.3	How to do Machine Learning in production	30
		8.4.4	Why machine learning in production is harder then you think	30
		8.4.5	How to convince people machine learning works	31
9	Hov	v to St	ore Data	33
	9.1		Modeling	34
		9.1.1	How to find out how you need to store data for the business case.	34
		9.1.2	How to decide what kind of storage you need to use	34

	9.2	SQL .		34				
		9.2.1	Database Design	34				
		9.2.2	SQL Queries	34				
		9.2.3	Stored Procedures	34				
		9.2.4	ODBC/JDBC Server Connections	34				
	9.3	NoSQI		34				
		9.3.1	KeyValue Stores (HBase)	34				
		9.3.2	Document Stores (HDFS, MongoDB)	34				
		9.3.3	Hive Warehouse	86				
		9.3.4	Impala	86				
		9.3.5	Time Series Databases (?)	36				
		9.3.6	MPP Databases (Greenplum)	37				
10	How	to Vi	sualize Data 3	7				
	10.1	Mobile	e Apps	37				
		10.1.1	Android & IOS basics	37				
		10.1.2	How to design APIs for mobile apps	37				
	10.2	How to	o use Webservers to display content	37				
		10.2.1	Tomcat	8				
		10.2.2	Jetty	8				
		10.2.3	NodeRED	8				
		10.2.4	React	8				
	10.3	Busine	ss Intelligence Tools	8				
		10.3.1	Tableau	8				
		10.3.2	PowerBI	8				
		10.3.3	Quliksense	8				
	10.4	Identit	y & Device Management	8				
		10.4.1	What is a digital twin?	8				
		10.4.2	Active Directory	8				
11	1 Case Studies 3							
	11.1	Data S	Science @Airbnb	8				
	11.2	Data S	Science @Netflix	8				
	11.3	Data S	Science @Uber	8				
	11.4	Data S	Sciecne @Zalando	88				
<b>12</b>	DEV	V/OPS	3	8				
	12.1	Hadoo	p	88				
		12.1.1	Hadoop Cluster setup and management with Cloudera Manager					
			(for example)	39				

12.1.2	Spark code from coding to production	36
12.1.3	How to monitor and manage data processing pipelines	39
12.1.4	Oozie	36
12.1.5	Airflow Application management	36
12.1.6	Creating Statistics with Spark and Kafka	39

## 1 Introduction

What do you actually need to learn to become an awesome data engineer? Look no further, you find it here.

How to use this document: This is not a training! It's a collection of skills, that I value highly in my daily work as a data engineer. It's intended to be a starting point for you to find the topics to look into.

This project is a work in progress! Over the next weeks I am going to share with you my thoughts on why each topic is important. I also try to include links to useful resources.

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- Andreas