

- 1. Basic Knowledge
  - 1.1 Linear Algebra
  - 1.2 Probability
- 2. Machine Learning
- 3. Algorithmic Machine Learning and Data Science
  - 3.1 The Power of Randomness
  - 3.2 First Order Optimization
  - 3.3 Spectral Algorithms
  - 3.5 Fourier Methods
- 4. Deep Learning
  - 4.1 NLP
  - 4.2 CV
  - 4.3 Generative Model
  - 4.4 Reinforcement Learning
- 5. Computer Vision
- 6. AI

(Updating...)

*This files summarizes some learning resources about ML/ DL/AI/ALML. Some sections here may have intersection, I mainly organize it according to the course I have taken.*

*Some recourses are from my course at NYU Tandon. Some are wonderful recourses on the internet.*

## 1. Basic Knowledge

### 1.1 Linear Algebra

A short review: [http://web.stanford.edu/class/cs246/handouts/CS246\\_LinAlg\\_review.pdf](http://web.stanford.edu/class/cs246/handouts/CS246_LinAlg_review.pdf)

Sandford summary of basic linear algebra: <http://cs229.stanford.edu/section/cs229-linalg.pdf>

Space understanding for Linear Algebra (from 3 blue 1 brown):

[https://www.youtube.com/playlist?list=PLZHQObOWTQDPD3MizzM2xVFitgF8hE\\_ab&app=desktop](https://www.youtube.com/playlist?list=PLZHQObOWTQDPD3MizzM2xVFitgF8hE_ab&app=desktop)

My note can be found here:

### 1.2 Probability

Stanford "Review of Probability Theory - Arian Maleki and Tome Do"<http://cs229.stanford.edu/section/cs229-prob.pdf>

## 2. Machine Learning

## 3. Algorithmic Machine Learning and Data Science

Wonderful courses:

CS168: The Modern Algorithmic Toolbox <https://web.stanford.edu/class/cs168/index.html>

CS246: Mining Massive Data Sets <http://web.stanford.edu/class/cs246/>

### 3.1 The Power of Randomness

<a href="https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec1.pdf">https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec1.pdf</a>	Typed notes for hashing - Christopher Musco
<a href="https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec10.pdf">https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec10.pdf</a>	Typed notes - Christopher Musco: Dimensionality reduction and the Johnson-Lindenstrauss Lemma
<a href="https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec12.pdf">https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec12.pdf</a>	Typed notes - Christopher Musco: Nearest Neighbor Search and Locality Sensitive Hashing

### 3.2 First Order Optimization

#### Reading:

Website	Description
<a href="https://www.offconvex.org/">https://www.offconvex.org/</a>	diverse, theory-learning blog which discusses lots of problems and approaches in understanding the optimization of Non-convex problems
<a href="https://ee227c.github.io/notes/ee227c-notes.pdf">https://ee227c.github.io/notes/ee227c-notes.pdf</a>	Course Notes for EE227C (Spring 2018): Convex Optimization and Approximation - Moritz Hardt's proofs of gradient descent convergence in all the regimes discussed in class.
<a href="https://arxiv.org/pdf/1909.05207.pdf">https://arxiv.org/pdf/1909.05207.pdf</a>	Introduction to Online Convex Optimization Elad Hazan's book,
<a href="https://arxiv.org/pdf/1405.4980.pdf">https://arxiv.org/pdf/1405.4980.pdf</a>	Convex Optimization: Algorithms and Complexity Sébastien Bubeck

Analyzing gradient descent and project gradient descent for convex problems

Online and stochastic gradient descent

Smoothness, strong convexity, conditioning

Preconditioning, acceleration, coordinate descent, non-convex optimization

### 3.3 Spectral Algorithms

<a href="https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec14.pdf">https://www.cs.princeton.edu/courses/archive/fall18/cos521/Lectures/lec14.pdf</a>	Typed notes - Christopher Musco: The Power Method and Spectral Methods for Graph Partitioning
<a href="http://web.stanford.edu/class/cs168/l/111.pdf">http://web.stanford.edu/class/cs168/l/111.pdf</a>	Stanford CS168: The Modern Algorithmic Toolbox Lectures #11: Spectral Graph Theory, I

## 3.5 Fourier Methods

## 4. Deep Learning

Open Course:

MIT Introduction to DeepLearning - Alexander Amini: <http://introtodeeplearning.com/>

Precise and short, covered all the topics, wonderful speaker

### 4.1 NLP

### 4.2 CV

### 4.3 Generative Model

### 4.4 Reinforcement Learning

good open course:

UC Berkeley CS 285 <http://rail.eecs.berkeley.edu/deeprlcourse/> Many other good courses/books can be found under recourses tab.

[https://www.youtube.com/playlist?list=PL\\_iWQOsE6TfURIIhCrlt-wj9ByIVpbfGc](https://www.youtube.com/playlist?list=PL_iWQOsE6TfURIIhCrlt-wj9ByIVpbfGc)

RL Course by David Silver - [https://www.youtube.com/watch?v=2pWv7GOvuf0&list=PLqYmG7hTraZBiG\\_XpjnPrSNw-1XQaM\\_gB](https://www.youtube.com/watch?v=2pWv7GOvuf0&list=PLqYmG7hTraZBiG_XpjnPrSNw-1XQaM_gB)

Blogs:	description
A (Long) Peek into Reinforcement Learning <a href="https://lilianweng.github.io/lil-log/2018/02/19/a-long-peek-into-reinforcement-learning.html#key-concepts">https://lilianweng.github.io/lil-log/2018/02/19/a-long-peek-into-reinforcement-learning.html#key-concepts</a>	
Policy Gradient Algorithms <a href="https://lilianweng.github.io/lil-log/2018/04/08/policy-gradient-algorithms.html">https://lilianweng.github.io/lil-log/2018/04/08/policy-gradient-algorithms.html</a>	

### Papers and discussions:

writer	paper	resource/review/explanation	description
Alex	Deep Reinforcement Learning Doesn't Work Yet <a href="https://www.alexirpan.com/2018/02/14/rl-hard.html">https://www.alexirpan.com/2018/02/14/rl-hard.html</a>	<a href="https://zhuanlan.zhihu.com/p/33936457">https://zhuanlan.zhihu.com/p/33936457</a>	some thing need to examine, for RL
David Pfau	Connecting Generative Adversarial Networks and Actor-Critic Methods <a href="https://arxiv.org/abs/1610.01945">https://arxiv.org/abs/1610.01945</a>	<a href="https://www.zhihu.com/question/60167306">https://www.zhihu.com/question/60167306</a>	

## **5. Computer Vision**

## **6. AI**

UCB - CS88 Introduction to AI <https://inst.eecs.berkeley.edu/~cs188/fa20/>