

# LECTURE 31

## CRASH COURSE ON MACHINE LEARNING

MCS 275 Spring 2022

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# LECTURE 31: CRASH COURSE ON MACHINE LEARNING

## Course bulletins:

- Watch video for lecture 30
- No in-person meeting on Friday - watch video for lecture 32 by end of Friday.
- Professor Dumas returns on Monday 4 April

**BEFORE I SAY ANYTHING...**

**WHAT DOES THE PHRASE  
"MACHINE LEARNING" MAKE YOU  
THINK OF?**

# CONTENTS

- What is machine learning?

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- What are some examples of machine learning?

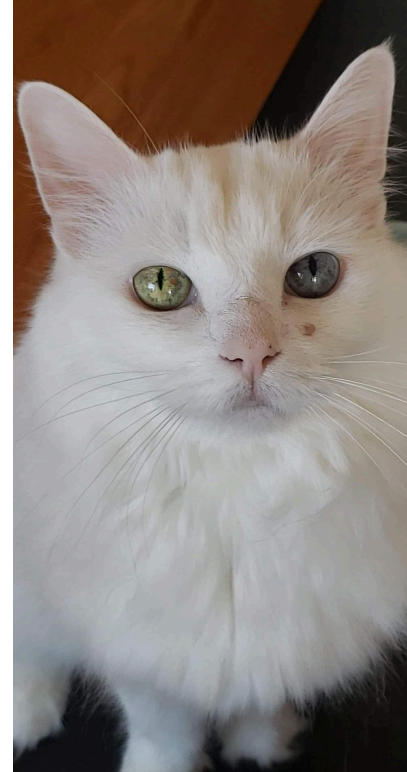
# CONTENTS

- What is machine learning?
- What are some examples of machine learning?
- How can you get into machine learning?

# WHAT IS MACHINE LEARNING?

"The study of computer algorithms that can improve automatically through experience and by the use of data." - [Wikipedia](#)

# MOTIVATION: DOG OR CAT?



How do we write a program to find out whether an image of is a dog or a cat?



# MAIN TYPES OF MACHINE LEARNING:

**Supervised learning — Data is labeled.**

**Unsupervised learning — Data is not labeled.**

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We'll focus on supervised and unsupervised learning.

# QUICK EXAMPLES (UNSUPERVISED):

- **Recommendation systems** (e.g. Google search, YouTube/TikTok suggestions, Amazon suggestions)



Search bar containing the text: what is machine |

Search suggestions:

- what is machine **learning**
- what is machine **gun kelly's real name**
- what is machine **gun kelly's net worth**
- what is machine **code**
- what is machine **language**
- what is machine **learning used for**
- what is machine **learning in data science**
- what is machine **gun kelly famous for**
- what is machine **gun kelly**
- what is machine **politics**

Buttons: Google Search, I'm Feeling Lucky

Report inappropriate predictions

# QUICK EXAMPLES (UNSUPERVISED):

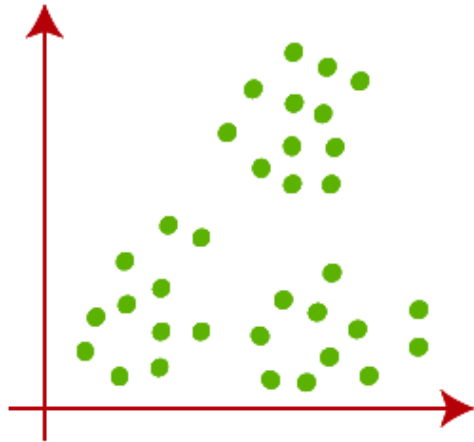
- **Dimensionality reduction** — Making data have fewer dimensions without losing its "meaning".

# QUICK EXAMPLES (UNSUPERVISED):

- **Clustering** — Finding suitable groups for unlabeled data.

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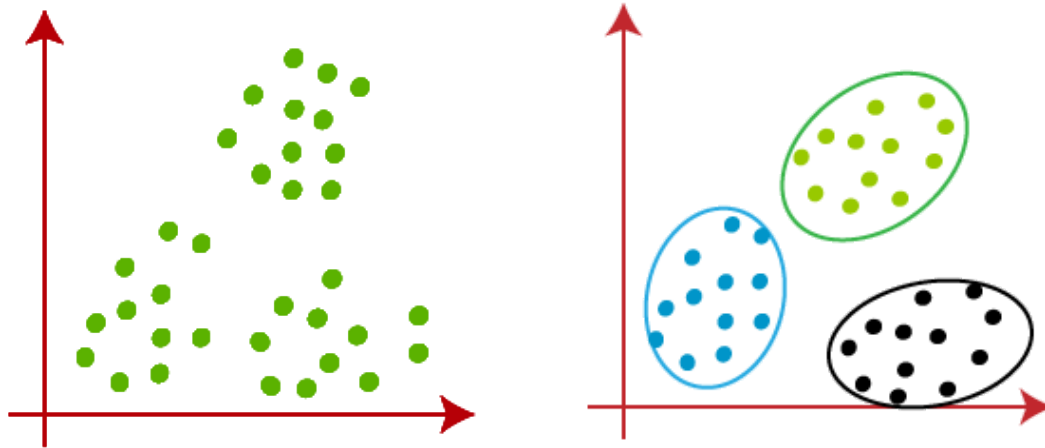
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**Classification** — Assigning labels to data with labeled categories. E.g. is an image a dog or a cat?



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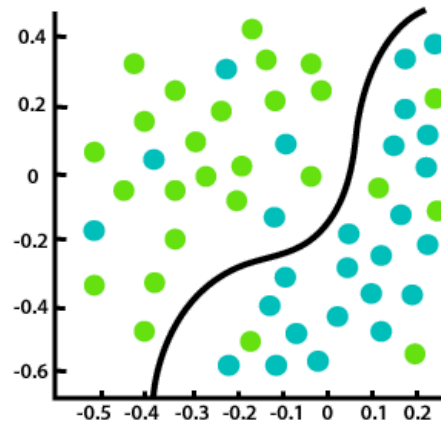
**Regression** — Assigning labels to data with continuous labels. If we know a student's midterm scores, can we predict their final scores?



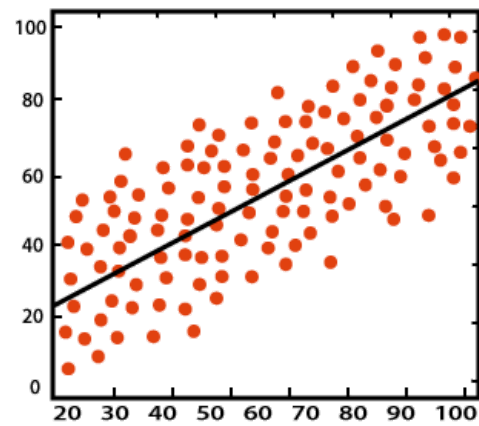
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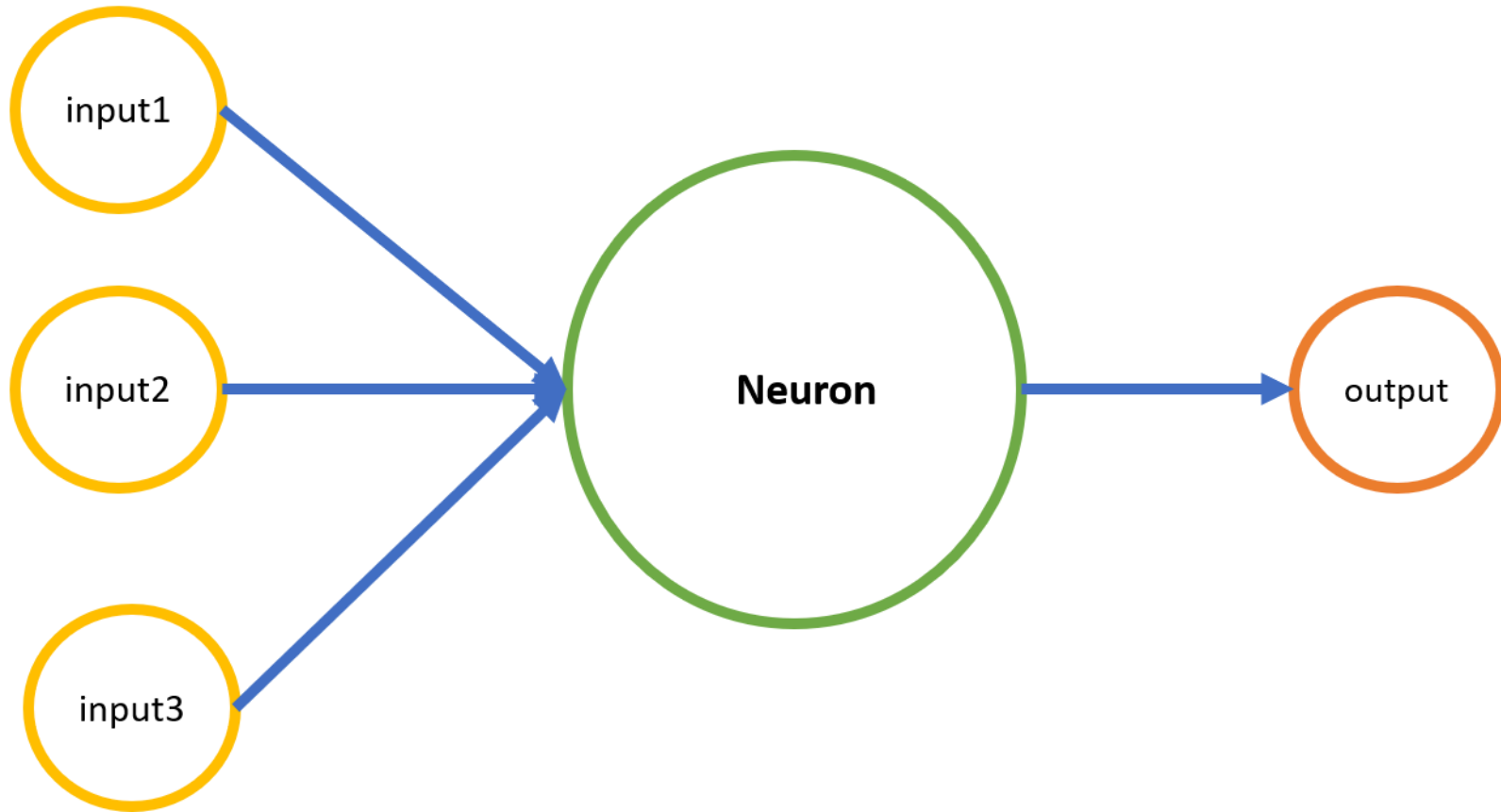


Classification



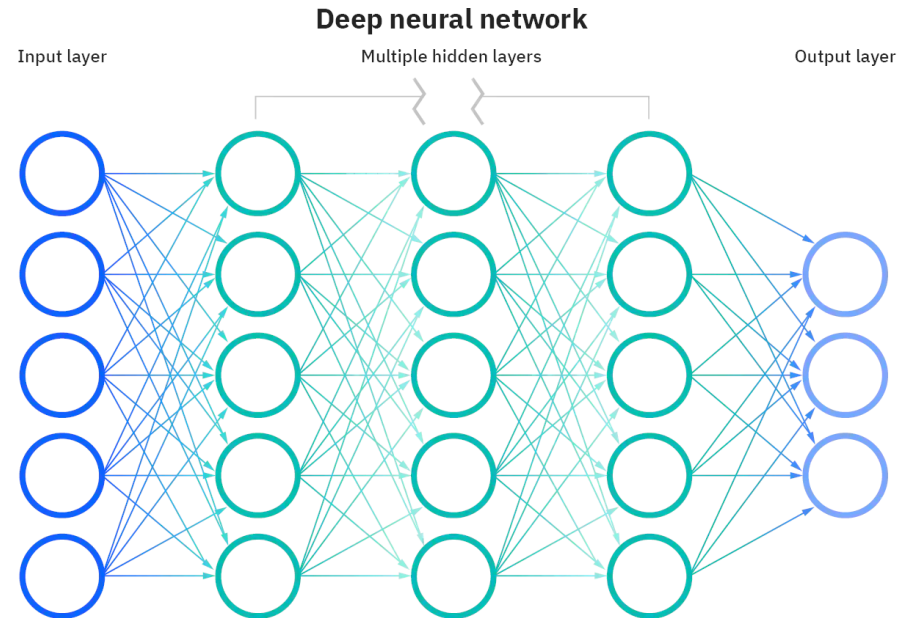
Regression

Let's define something called a "neuron":





# NEURAL NETWORKS



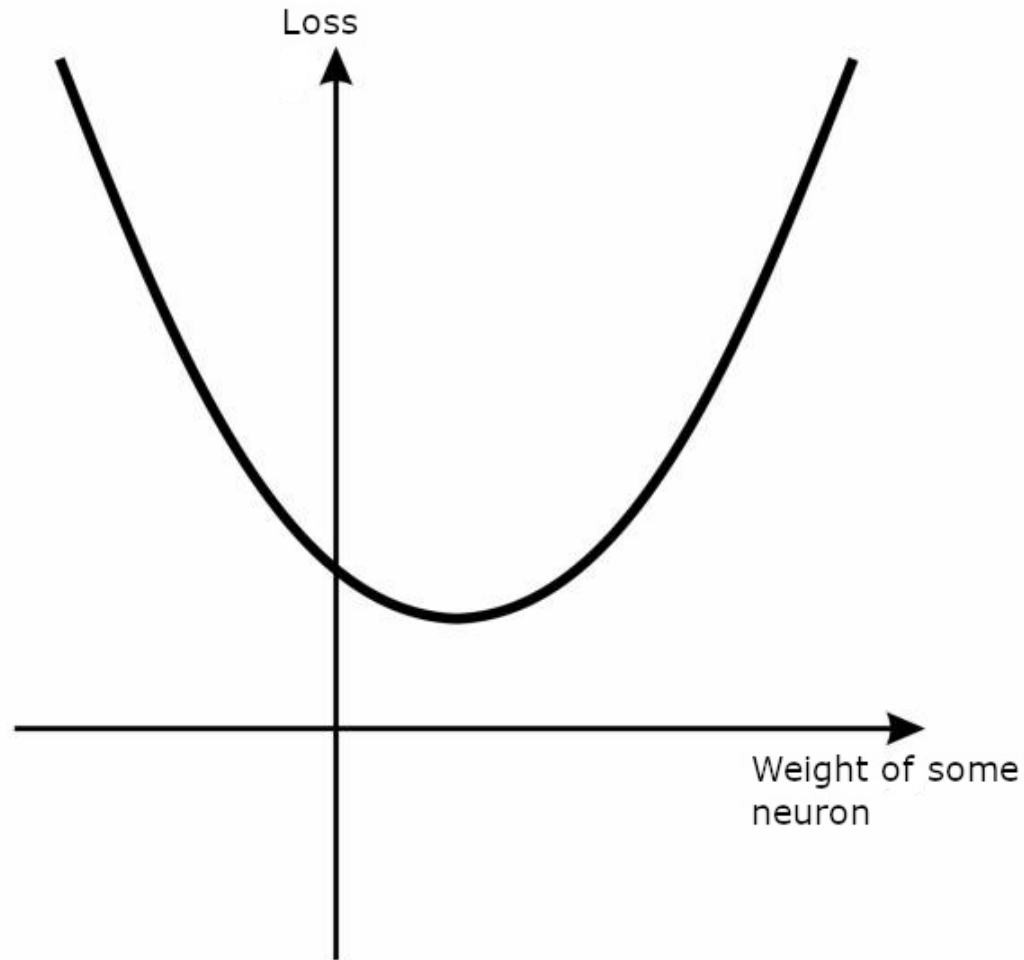
Source: [IBM](#)

They are a huge collection of "neurons", inspired by the brain.

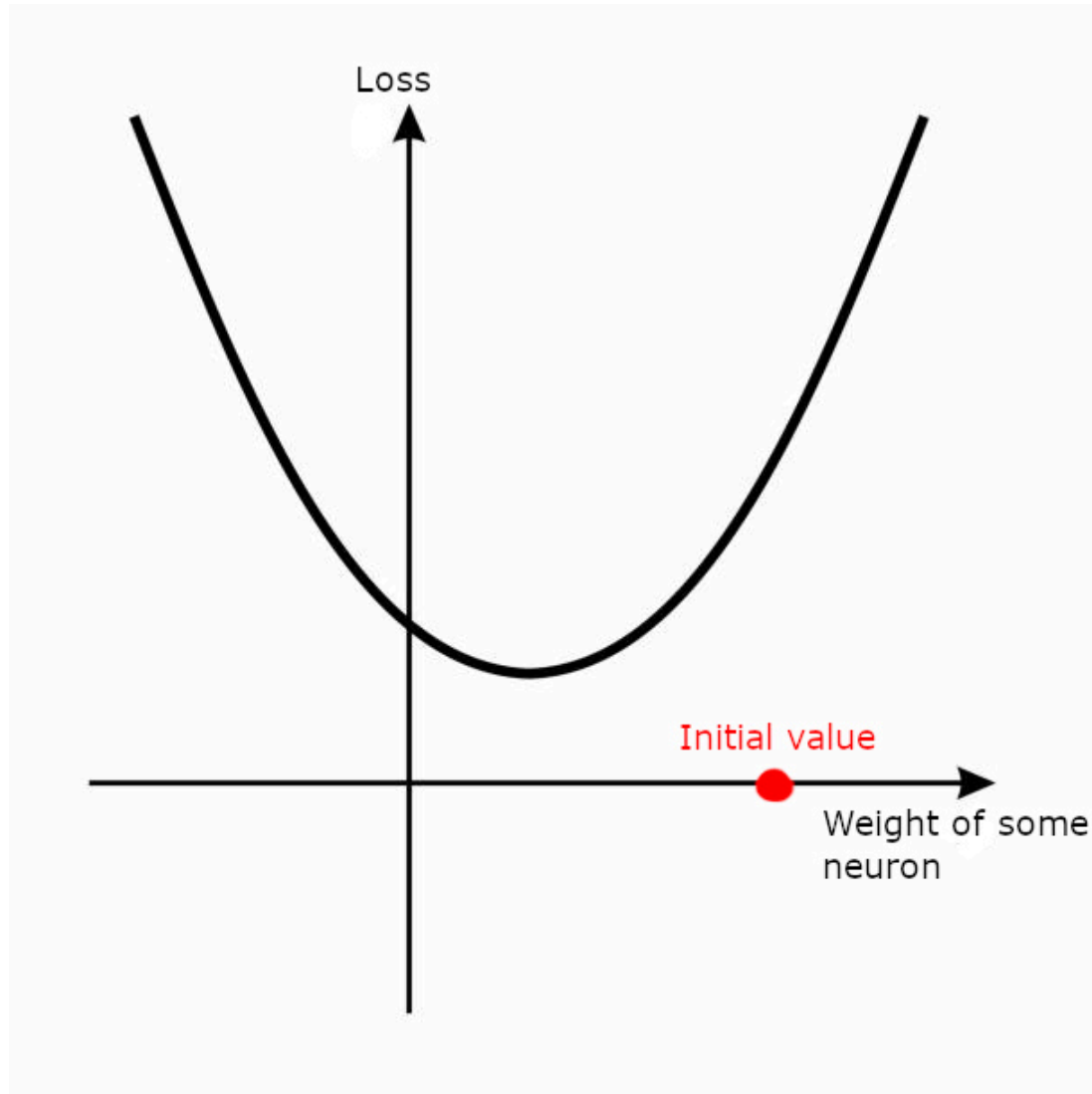
# LOSS FUNCTION

- We can have a function to characterize how "far" our model's prediction is from the true value.
- Then we can use calculus to find the values of the weight that make this as small as possible.

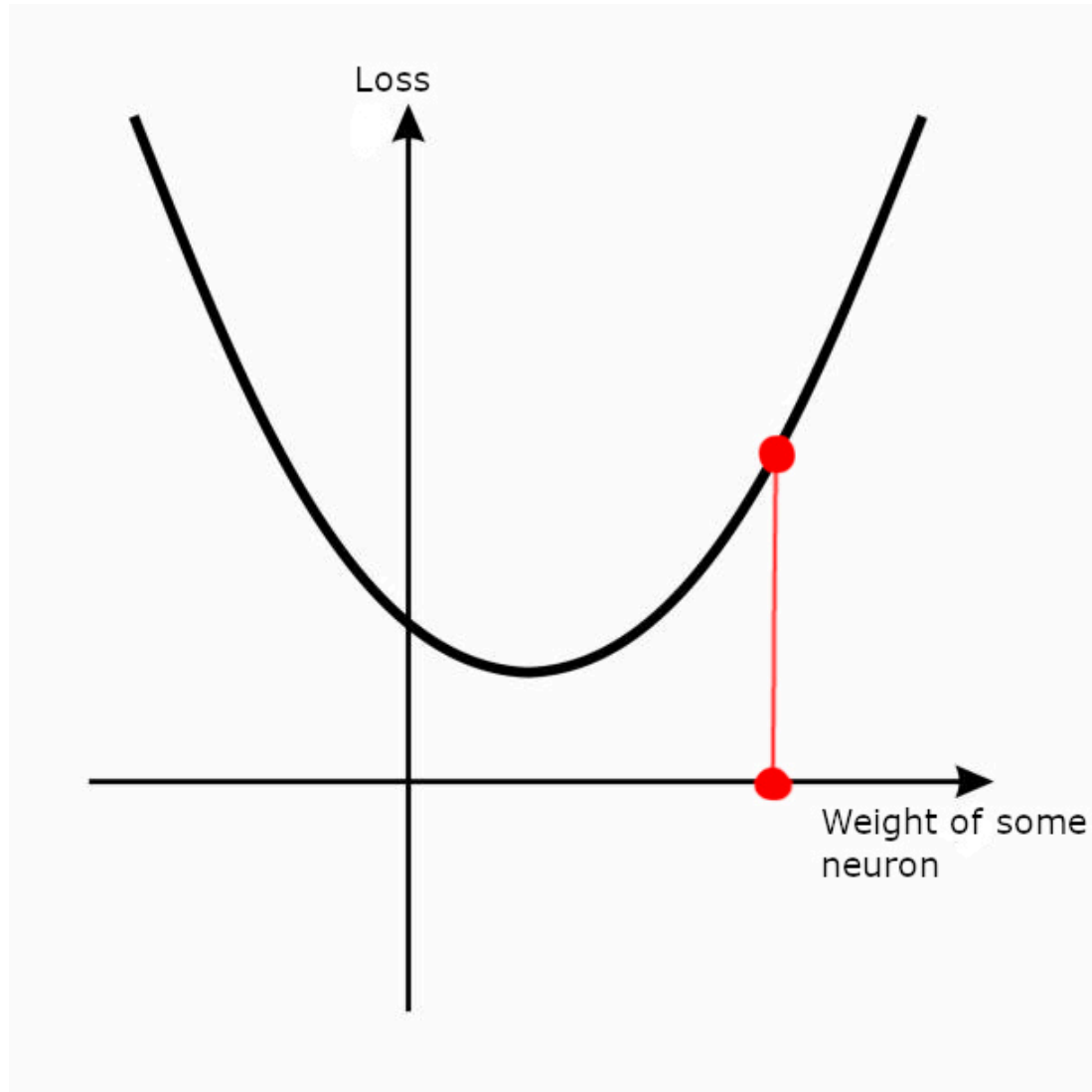
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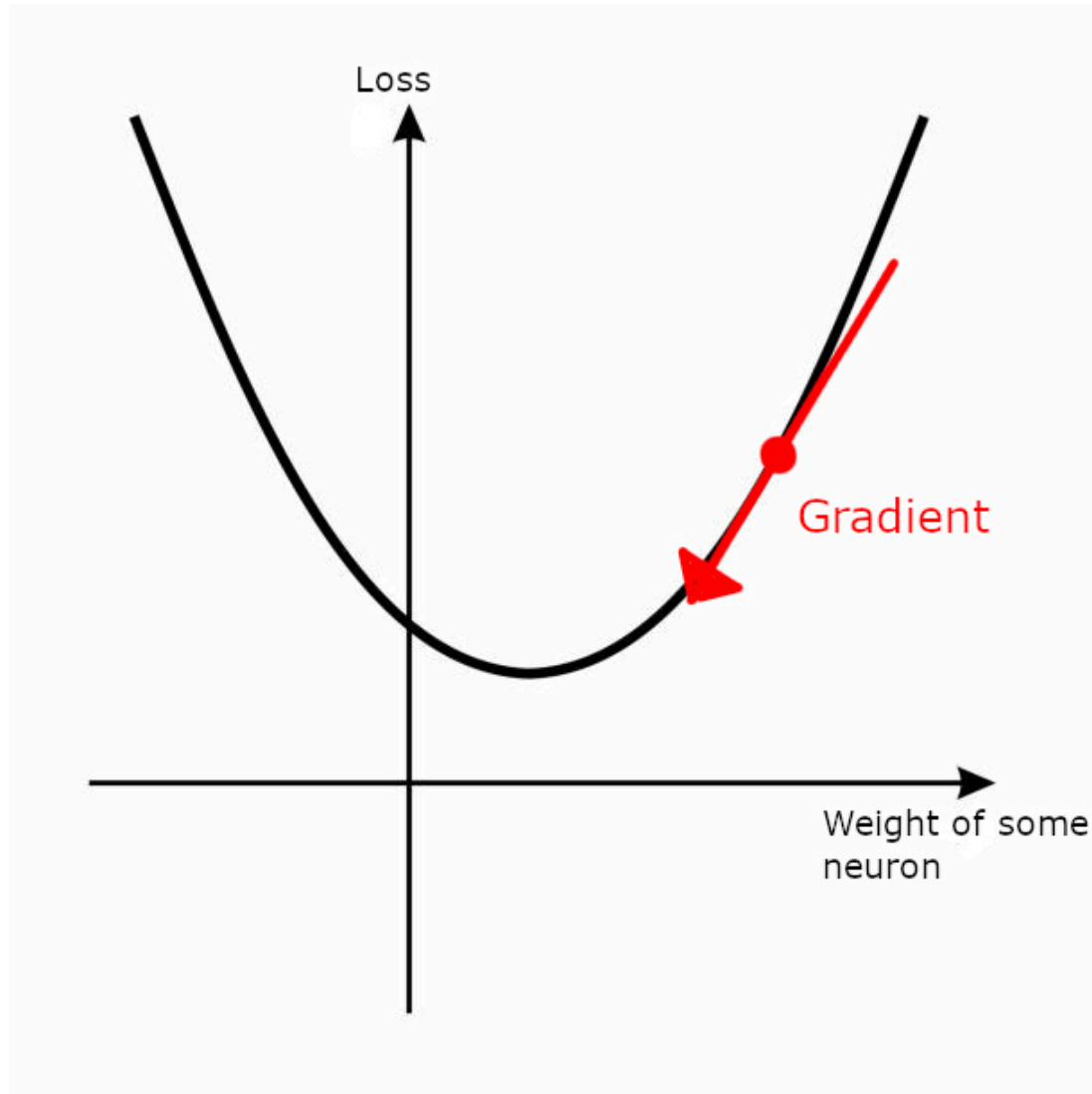
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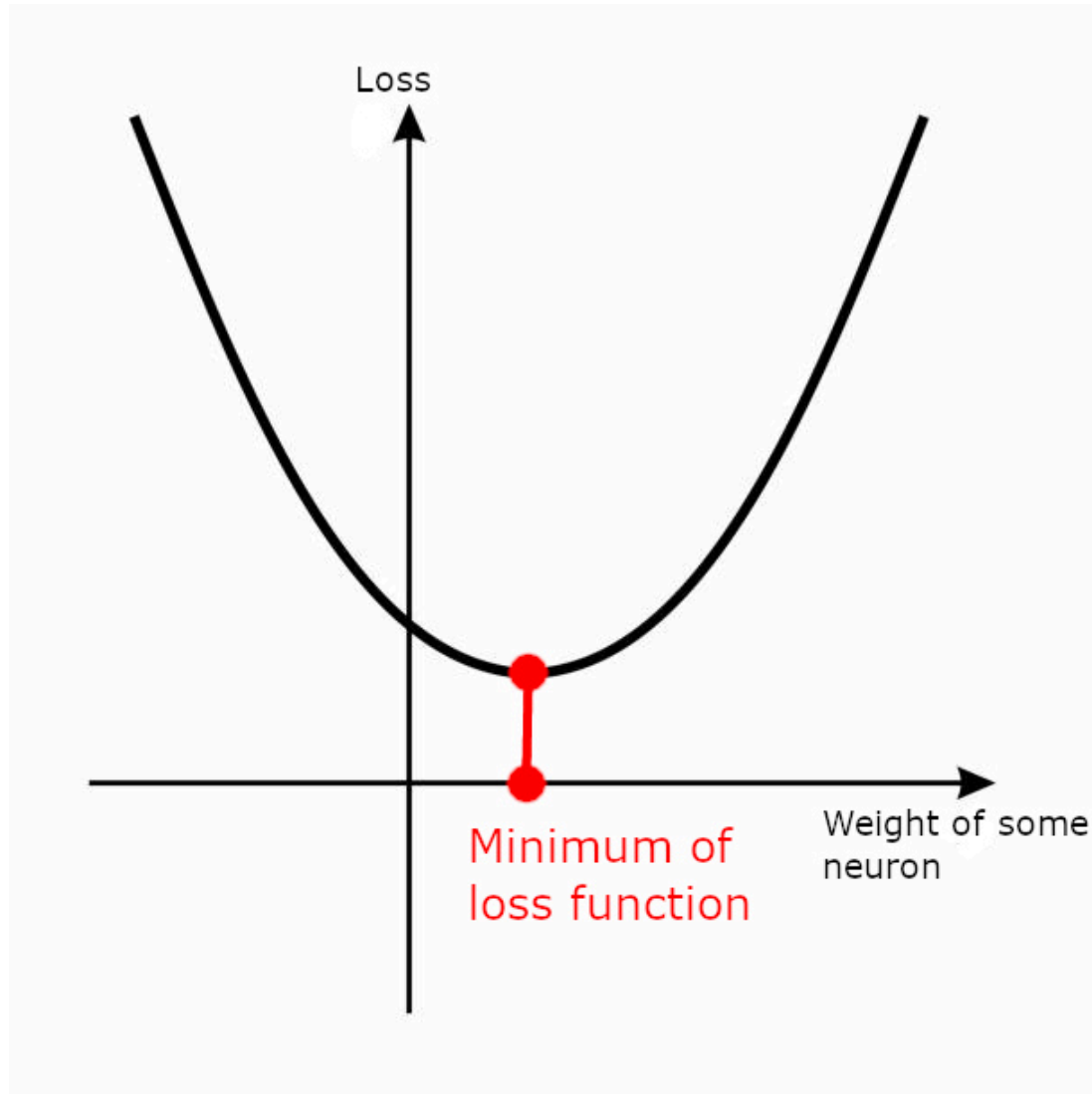
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# LET'S TRY IT

→ Jupyter notebook ←




# HOW CAN YOU DO MACHINE LEARNING WITH PYTHON?

- **scikit-learn** (simplest) — Good for general machine learning or quick implementations.
- **PyTorch** — Good for deep learning.
- **TensorFlow** — Good for deep learning.

# WHAT TOOLS HELP WITH MACHINE LEARNING?

- Python programming
- Linear algebra
- Calculus
- Statistics

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# CLASSES AT UIC RELATING TO MACHINE LEARNING

- **MCS 548** — Mathematical Theory of Artificial Intelligence,
- **CS 411** — Artificial Intelligence I,
- **CS 412** — Introduction to Machine Learning

# SUMMARY

- We saw the main types of machine learning (supervised, unsupervised, and reinforcement)
- We saw brief examples of supervised/unsupervised learning
- We saw neural networks and built one in scikit-learn

# INTERESTING RELATED LINKS

- 3Blue1Brown on YouTube: [video series on neural networks](#)
- Towards Data Science: [Types of machine learning](#)
- Machine Learning Mastery: [Your first machine learning project in Python step-by-step](#)

