

COSC 325: Introduction to Machine Learning

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Dr. Santos



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Lecture 02: Machine Learning Motivation



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**If you are not
registered in this
section, please
yield your seat to
registered
students.**



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Are you in the right place?

- Course

- Course: COSC 325, Introduction to Machine Learning
- Pre-Requisites: ECE 313 or ECE 317 or MATH 323; and MATH 251 or MATH 257 with a grade of C or better.
- Recommended Background: Python Programming Language, Numpy, SciKit Learn

- Logistics

- Location: MKB 524
- Lectures: Tuesday/Thursday: 11:20 am to 12:35 pm
- Canvas Link: <https://utk.instructure.com/courses/206990>

Discord Server
UTK Fall-24 COSC 325

Join @
<https://discord.gg/DXpnvT9R>

Class Announcements

Homework:

- First homework available in Canvas

Exams:

Lectures:

Join Discord Server:

UTK Fall-24 COSC 325

<https://discord.gg/DXpnvT9R>

Quizzes:

- The first quiz is due on Sunday.

Course Project:

- Instruction files will be available early next week.
- Form teams by 08/29. Otherwise, we will randomly assign you a team.
- Send your team details via Discord #team-creation channel.
- Students working in teams of two may be paired with a third student.

Today's Topics

What is Machine Learning?



Notation, Applications, and Tools



Last Lecture

- Course overview
- Work hard and have fun
- We are here to learn



What is Machine Learning?

“Machine learning is the field of study that gives computers the ability to learn without being explicitly programmed.”

— *Arthur L. Samuel, AI pioneer, 1959*



“...machine learning is a subcategory within the field of computer science which allows you to implement artificial intelligence. So, it’s kind of a mechanism to get you to artificial intelligence.”

—*Rana el Kaliouby, CEO at Affectiva*

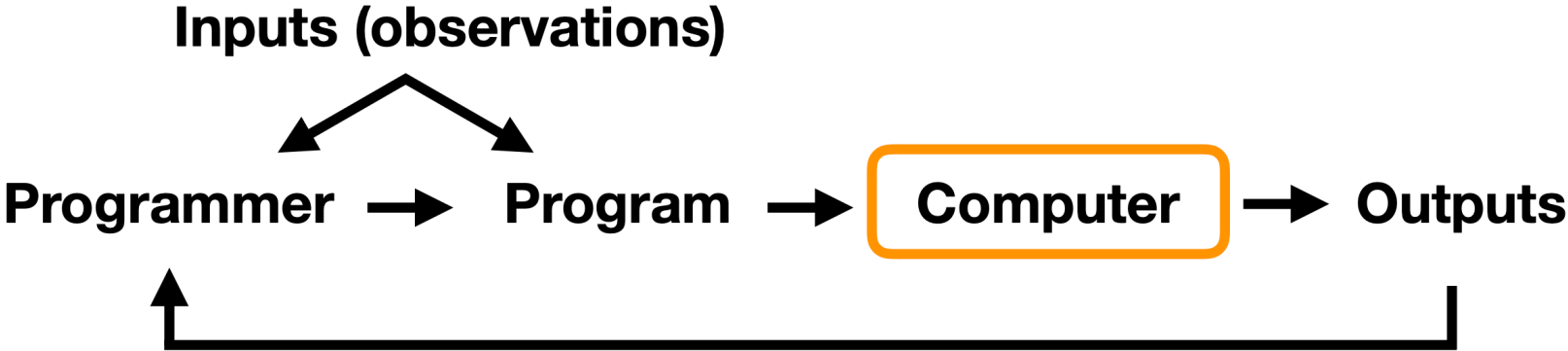


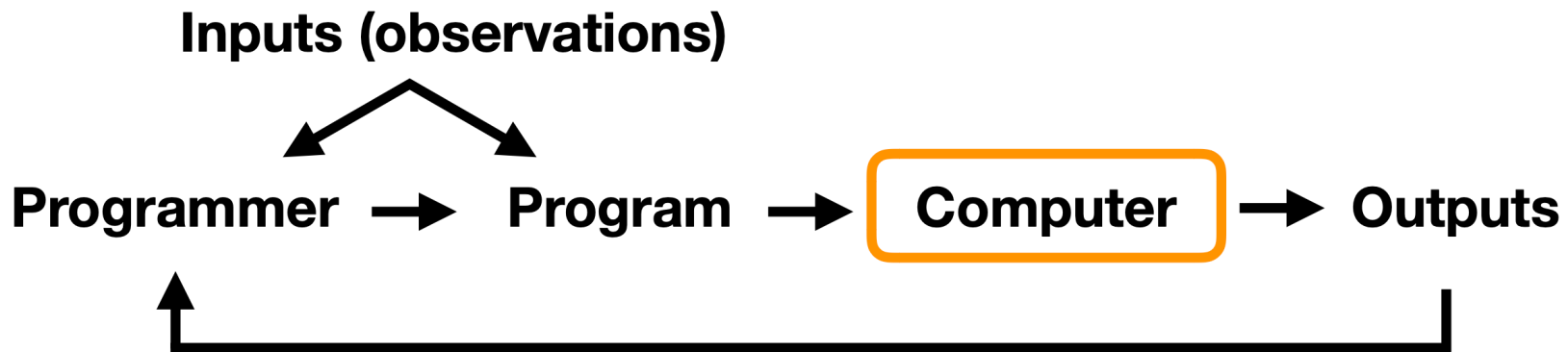
Machine Learning vs Computer Programming

- Traditional Programming:
 - Algorithms are sequences of instructions that are carried out to transform an input into an output
 - Fundamentally, they are lists of instructions
- Machine Learning:
 - The list of instructions is ***Learned*** from data
 - Useful when the sequence of instructions is difficult to define
 - Examples
 - Facial recognition
 - Autonomous driving



Traditional Programming Paradigm





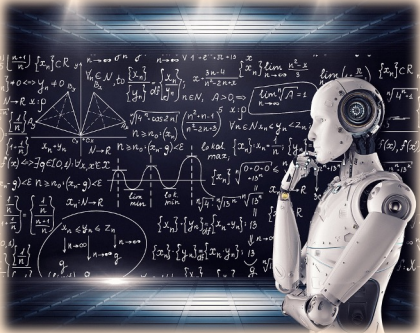
“Machine learning is the field of study that gives computers the ability to learn without being explicitly programmed.”
— *Arthur L. Samuel, AI pioneer, 1959*



AI vs Machine Learning

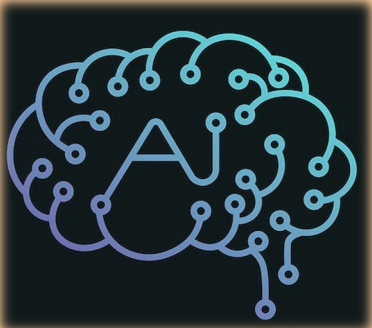
Artificial General Intelligence

Computers “mimic” how humans learn.



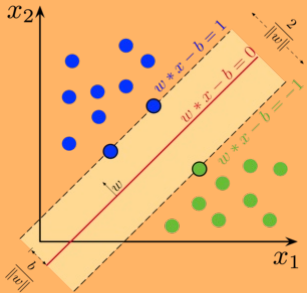
Artificial Intelligence

Computers mimic human behavior.



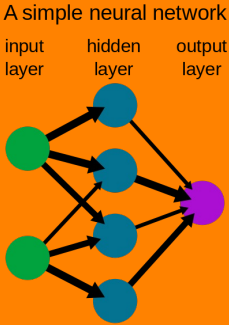
Machine Learning

Ability to learn without explicit hand-made rules.



Deep Learning

Automated extraction of patterns/features from raw data using multi-layer neural networks.



Teaching computers how to learn a task directly from data.

Figure inspired on MIT 6.S191 course slide.

Memory vs Learning

- What is an example of something that you memorize?
- What is an example of something that you learn?

Memory vs Learning

- What is an example of something that you memorize?
- What is an example of something that you learn?

1520 Middle Drive, Knoxville,
Tennessee 37796



Generalization

- A key component of “learning” is the ability to generalize
 - Take information that has been learned previously and apply it to new but related scenarios
- For a technique to be considered a machine learning approach, it **must** be able to generalize
- Thus, we must evaluate its ability to generalize



Memory vs Learning

- What is an example of something that you memorize?
- What is an example of something that you learn?

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Tennessee 37796

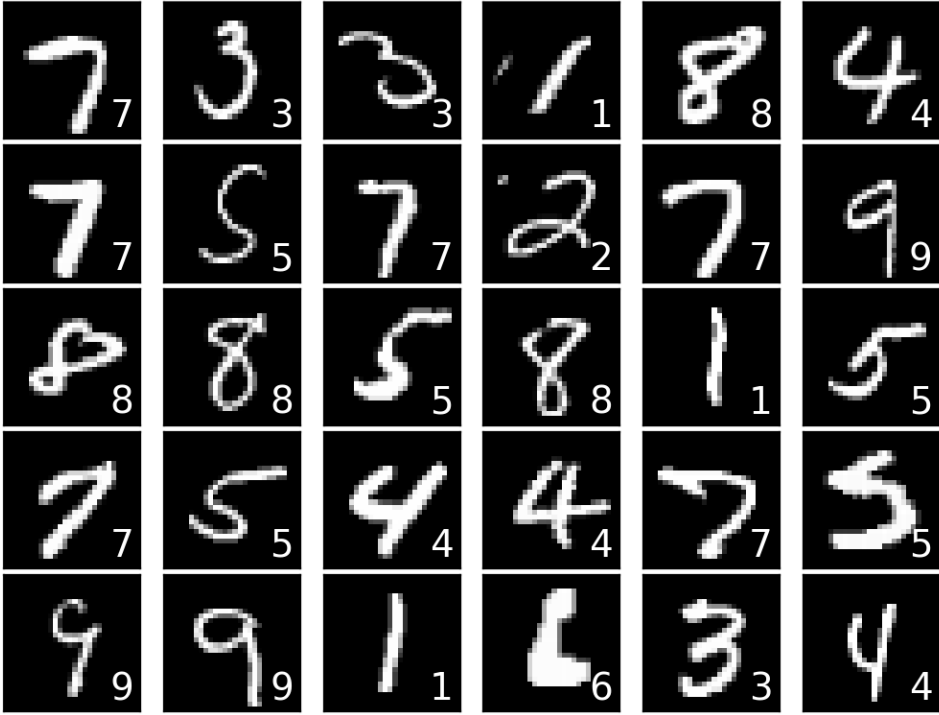


“A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P , if its performance at tasks in T , as measured by P , improves with experience E .” — **Tom Mitchell, Professor at Carnegie Mellon University**



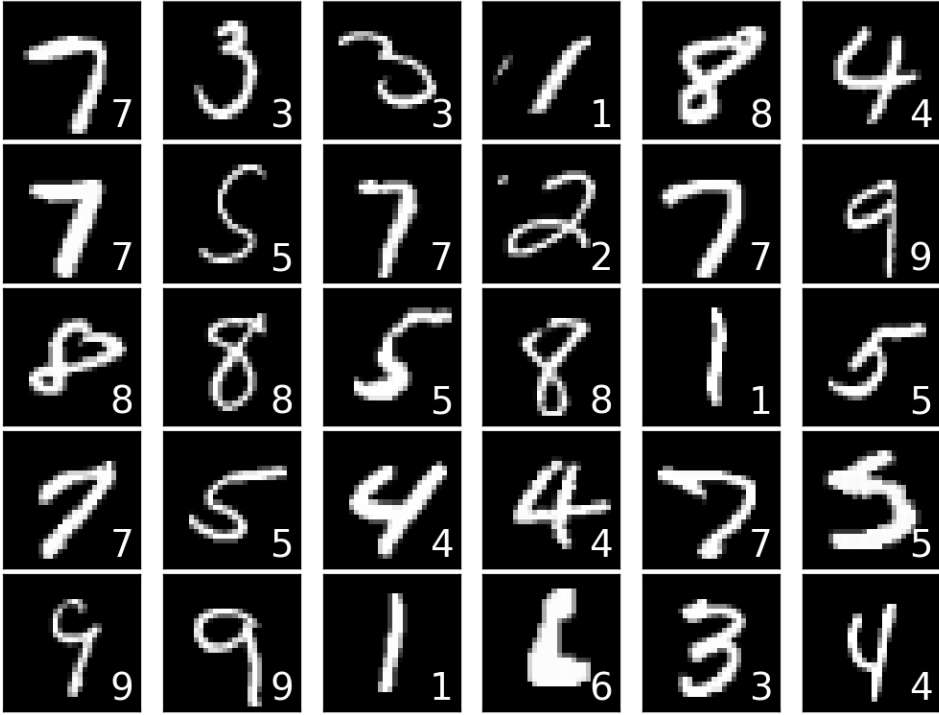
MNIST Images

- Task T : _____
- Performance measure P : _____
- Training experience E : _____



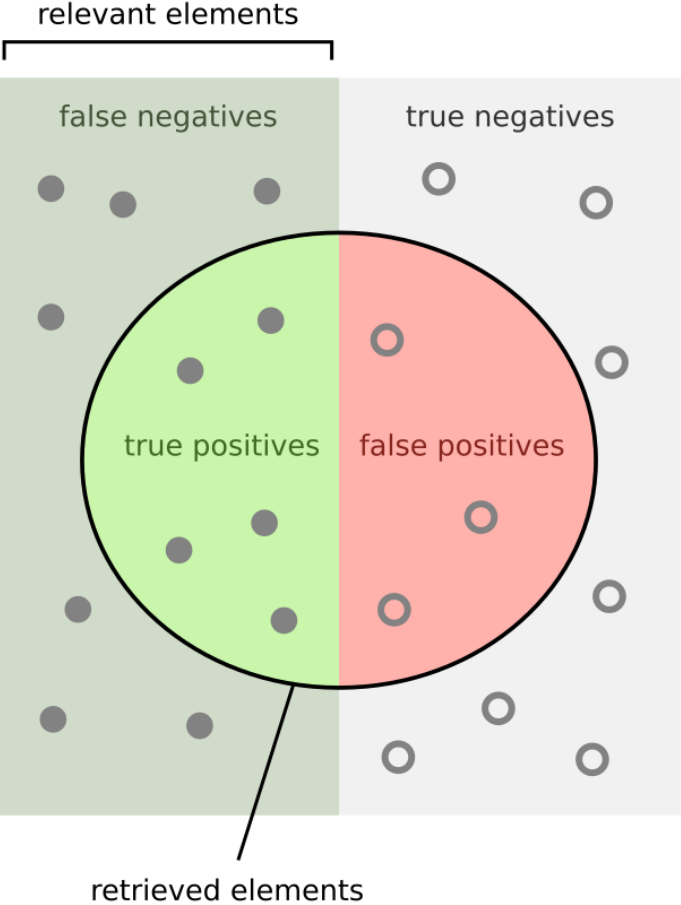
MNIST Images

- Task T : _____
- Performance measure P : _____
- Training experience E : _____



Features

Performance Metrics



How many retrieved items are relevant?

$$\text{Precision} = \frac{\text{true positives}}{\text{true positives} + \text{false positives}}$$

How many relevant items are retrieved?

$$\text{Recall} = \frac{\text{true positives}}{\text{true positives} + \text{false negatives}}$$

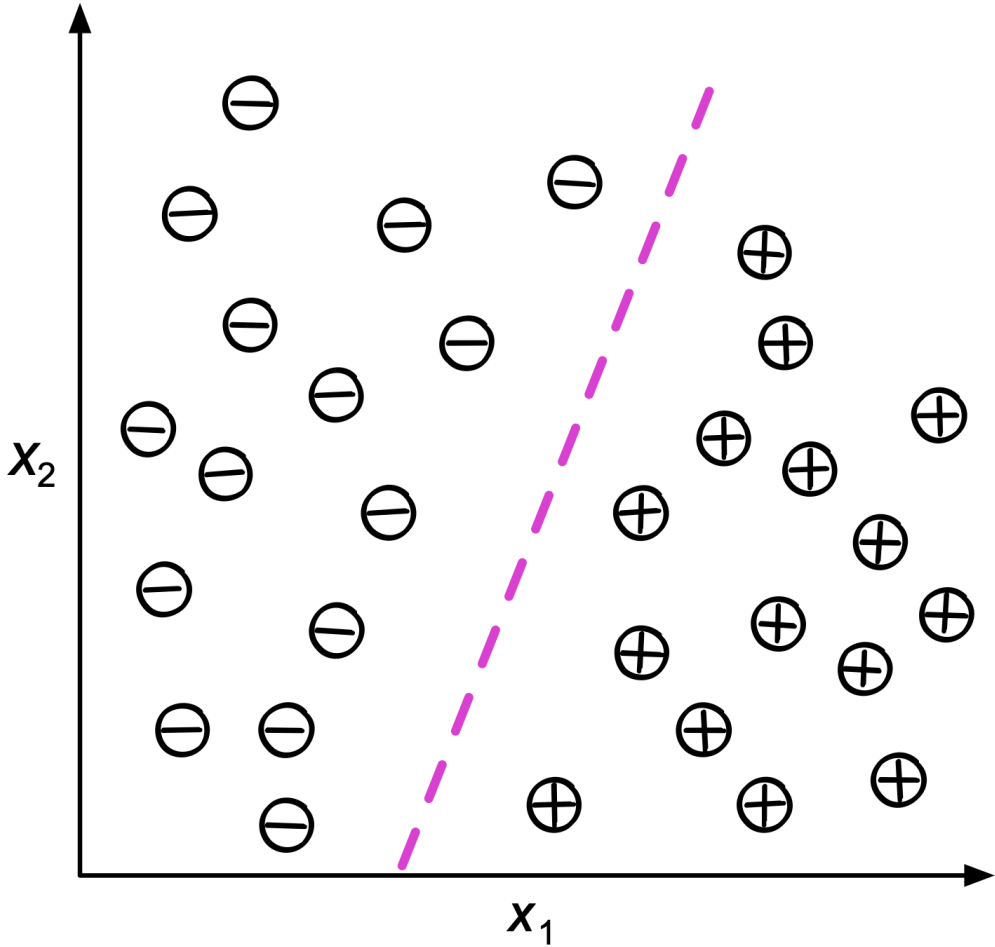
Categories of Machine Learning

Categories of Machine Learning

Supervised Learning

- > Labeled data
- > Direct feedback
- > Predict outcome/future

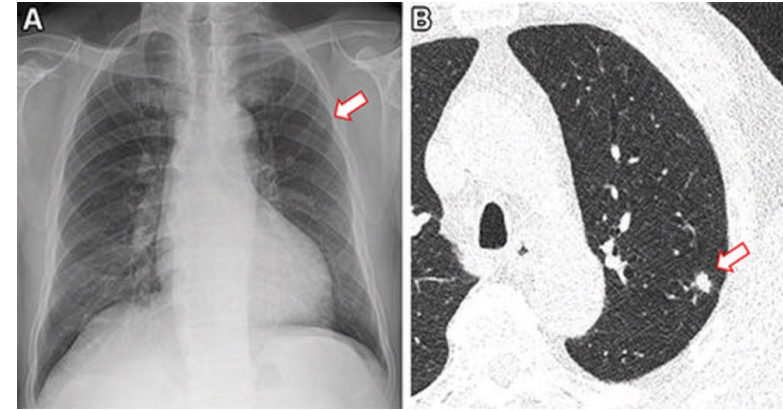
Supervised Learning: Classification



Binary Classification

Determine whether a tumor is benign or malign

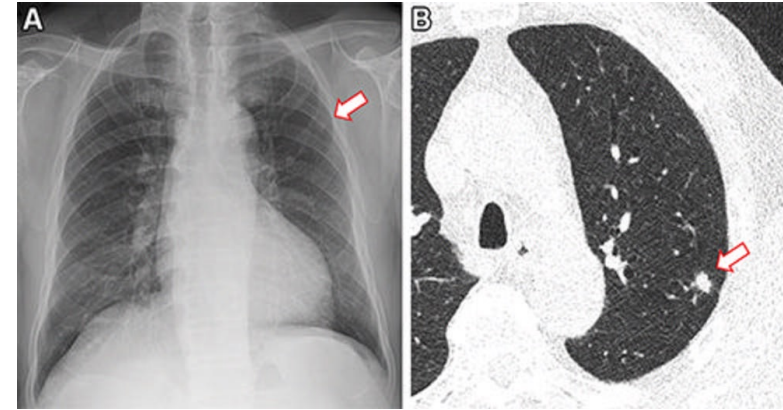
- Task:
- Features/Experience:
- Performance:



Binary Classification

Determine whether a tumor is benign or malign

- Task: Yes/No Predictions
- Features/Experience: Radius (mean of distances from the center to points on the perimeter), texture (standard deviation of gray-scale values), perimeter, area, smoothness (local variation in radius lengths), compactness ($\text{perimeter}^2 / \text{area} - 1.0$), concavity (severity of concave portions of the contour), concave points (number of concave portions of the contour), symmetry, fractal dimension.
- Performance: Accuracy, Binary Cross Entropy, Precision, Recall



Multi-Class Classification

Assign object label name to image

- Task:
- Features/Experience:
- Performance:

CIFAR10

airplane



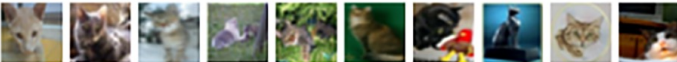
automobile



bird



cat



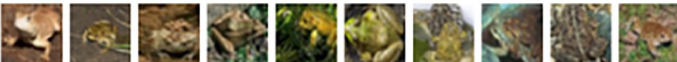
deer



dog



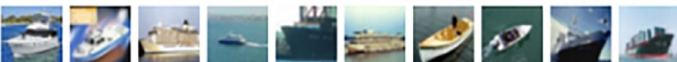
frog



horse



ship



truck



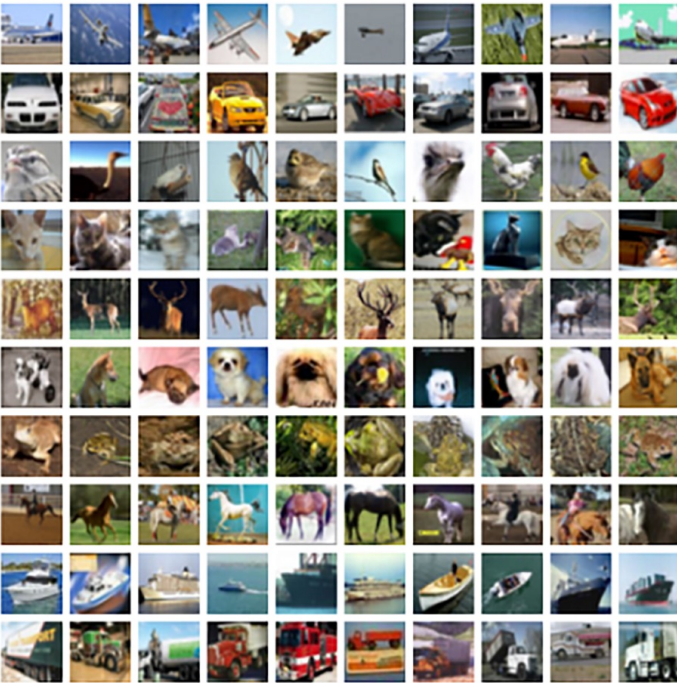
Multi-Class Classification

Assign object label name to image

- Task: Select a label within a set of possible predictions (labels)
- Features/Experience: *pixel values**, color histogram, edge angle histogram, correlation filter response
- Performance: Precision, Recall, Accuracy, F1 Score

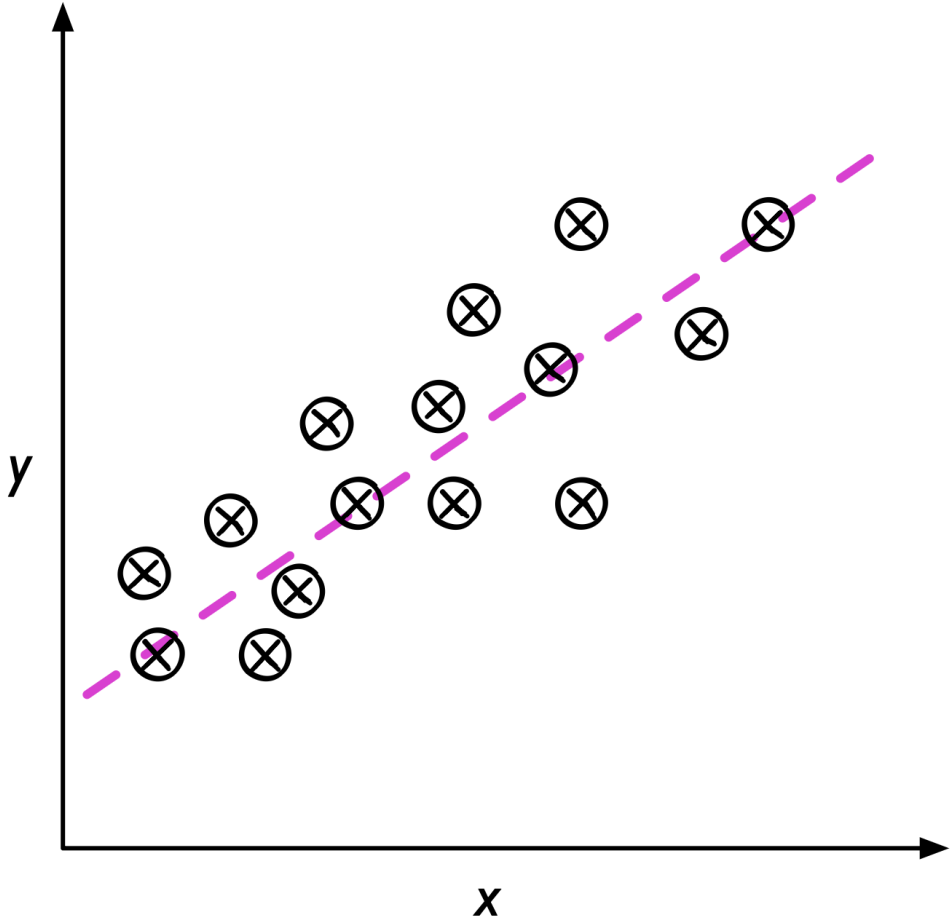
CIFAR10

airplane
automobile
bird
cat
deer
dog
frog
horse
ship
truck



**NNs*

Supervised Learning: Regression



Examples:

- Market value
- Time to failure
- Age
- Weight
- Size
- Location

Regression Example

Quantitative measure of diabetes disease progression one year after baseline.

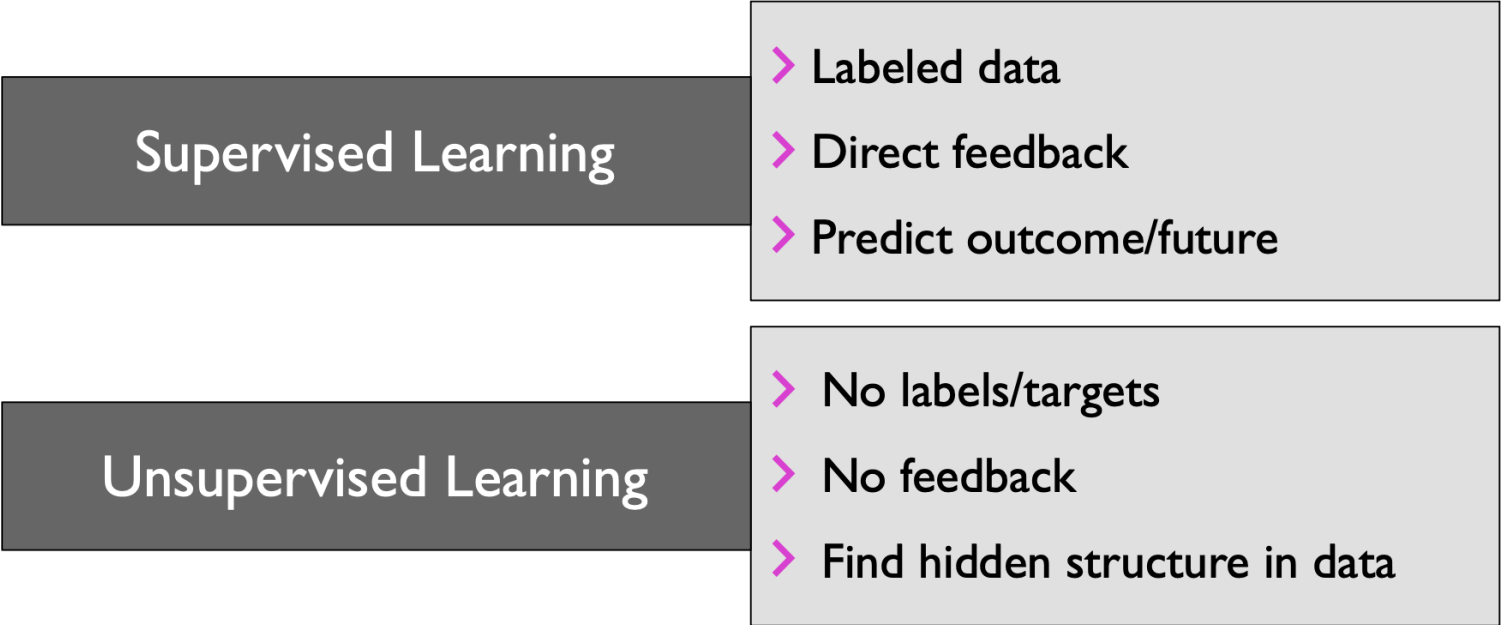
- Task:
- Features/Experience:
- Performance:

Regression Example

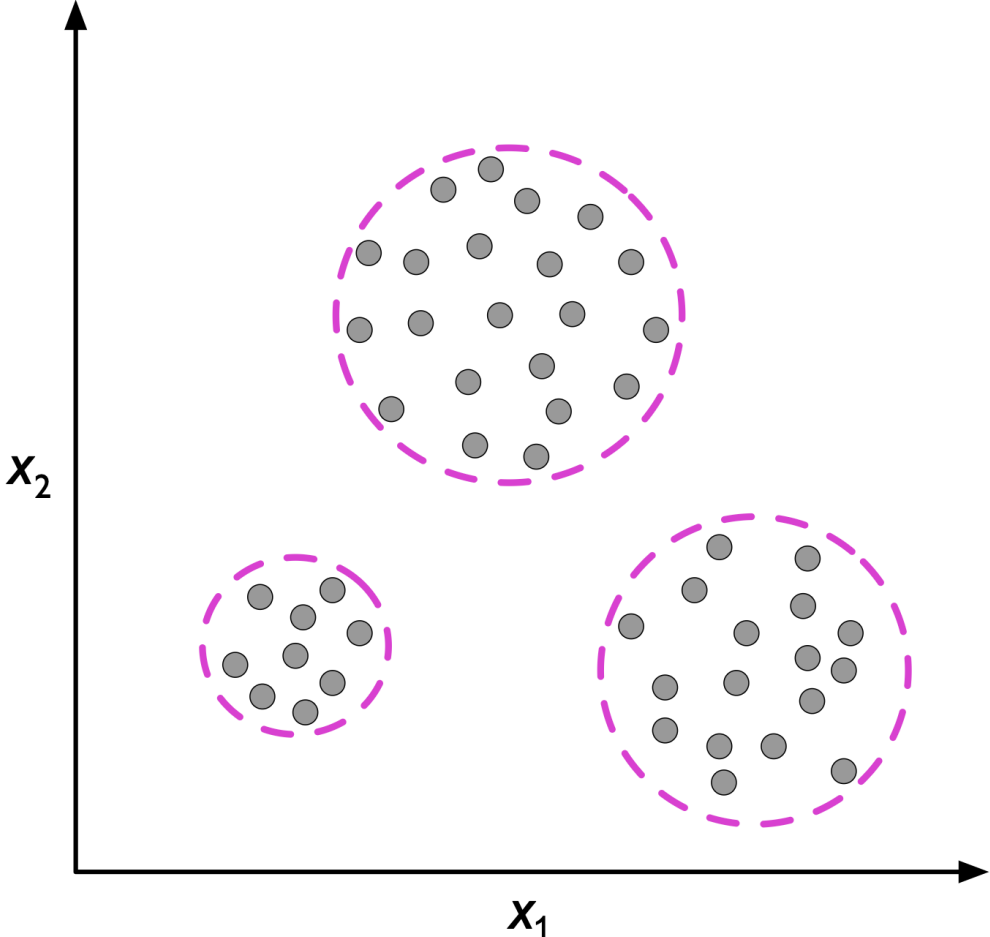
Quantitative measure of diabetes disease progression one year after baseline.

- Task: Predict a real value
- Features/Experience: Age in years, sex, BMI (body mass index), average blood pressure, total serum cholesterol, low-density lipoproteins, high-density lipoproteins, total cholesterol / HDL, possibly log of serum triglycerides level, and blood sugar level
- Performance: MSE between predicted and true values

Categories of Machine Learning



Unsupervised Learning: Clustering



Ranking Example

Return the top 10 most similar movies to a query movie

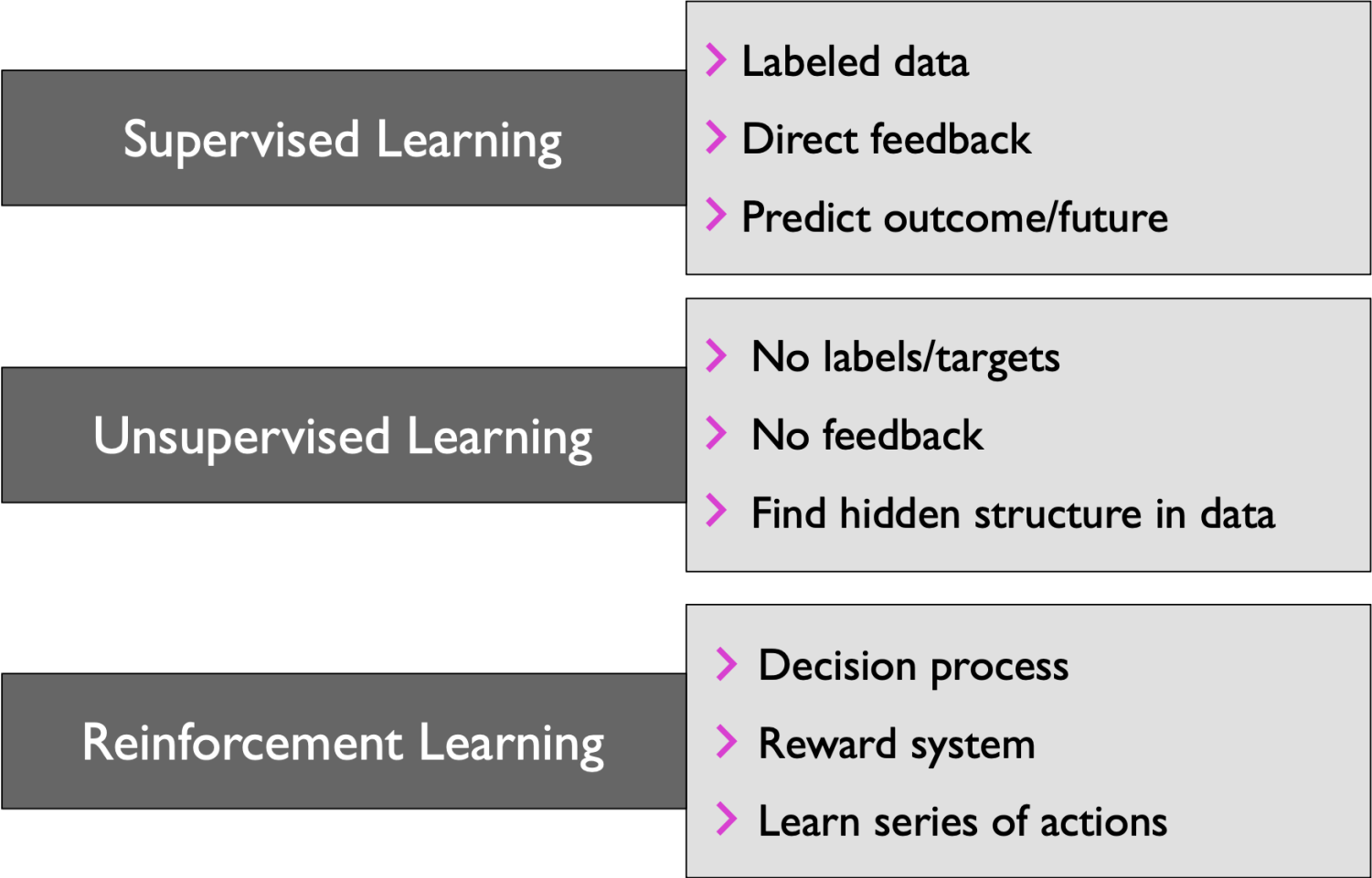
- Task: Compute relevance list
- Features/Experience:
- Performance:

Ranking Example

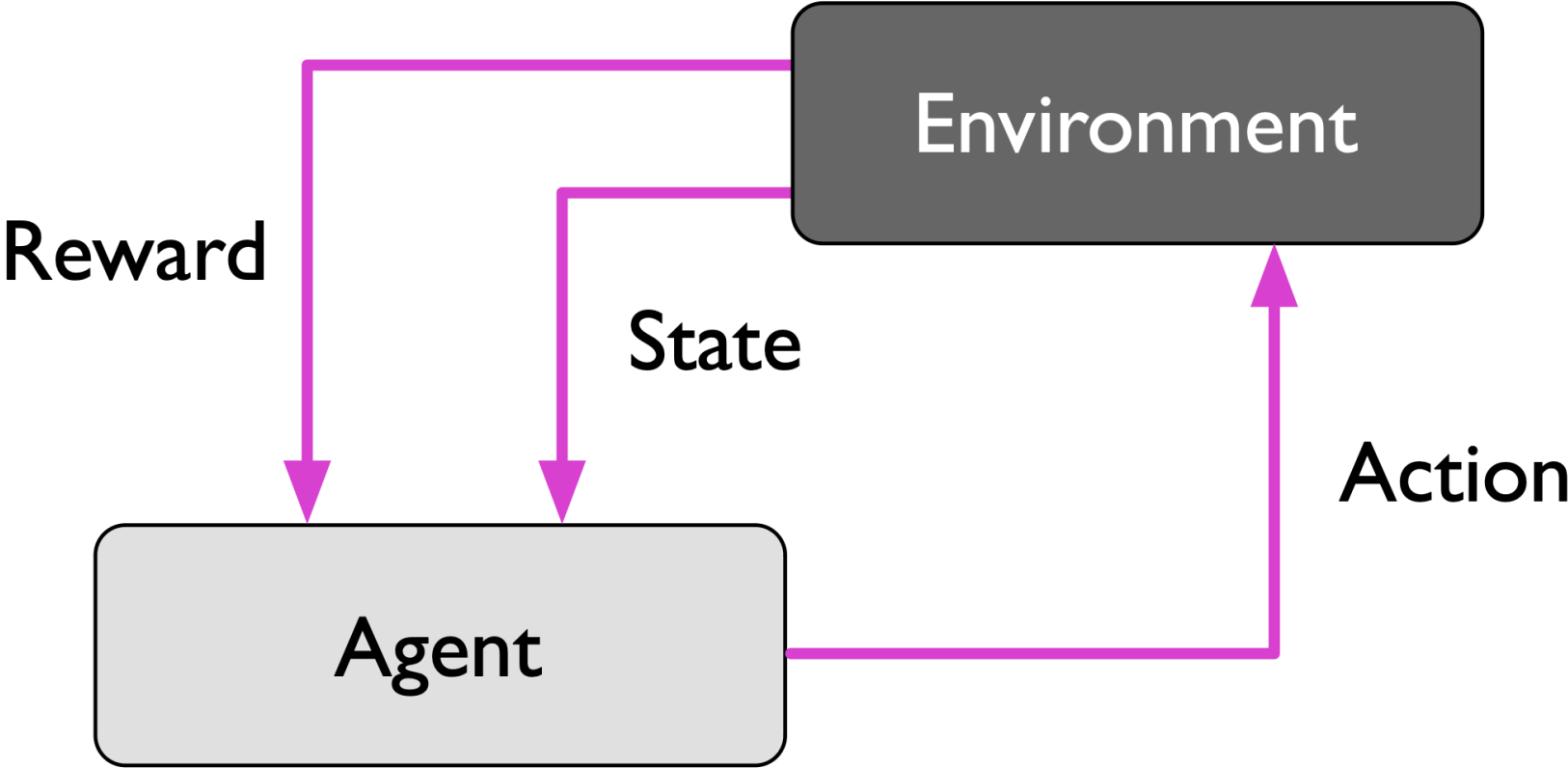
Return the top 10 most similar movies to a query movie

- Task: Compute relevance list
- Features/Experience: title, genre, year, synopsis, language, director, starring actors, studio, content advisory, maturity rating, [customer watch history].
- Performance: Recall@k from past user behavior

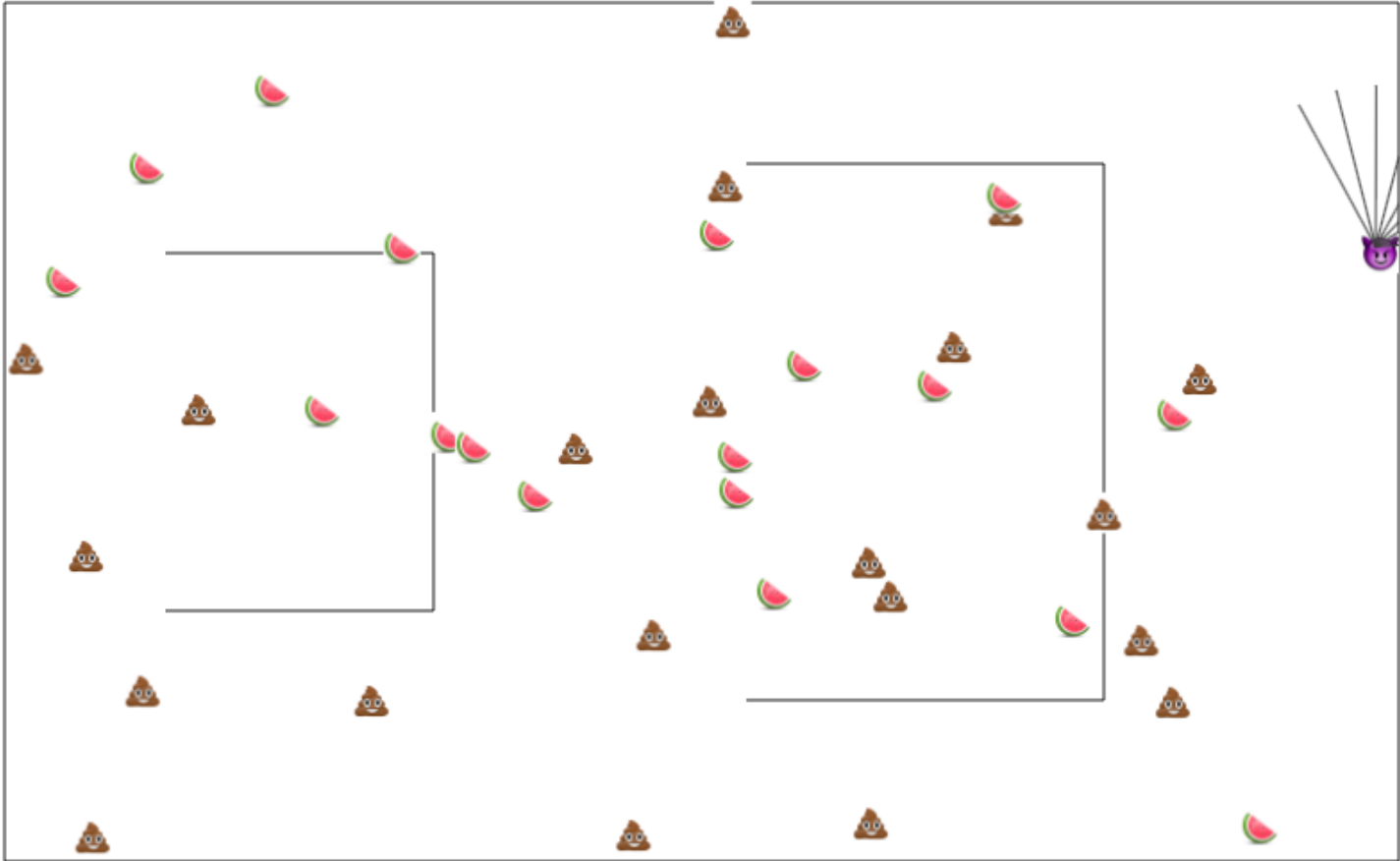
Categories of Machine Learning



Reinforcement Learning



Example



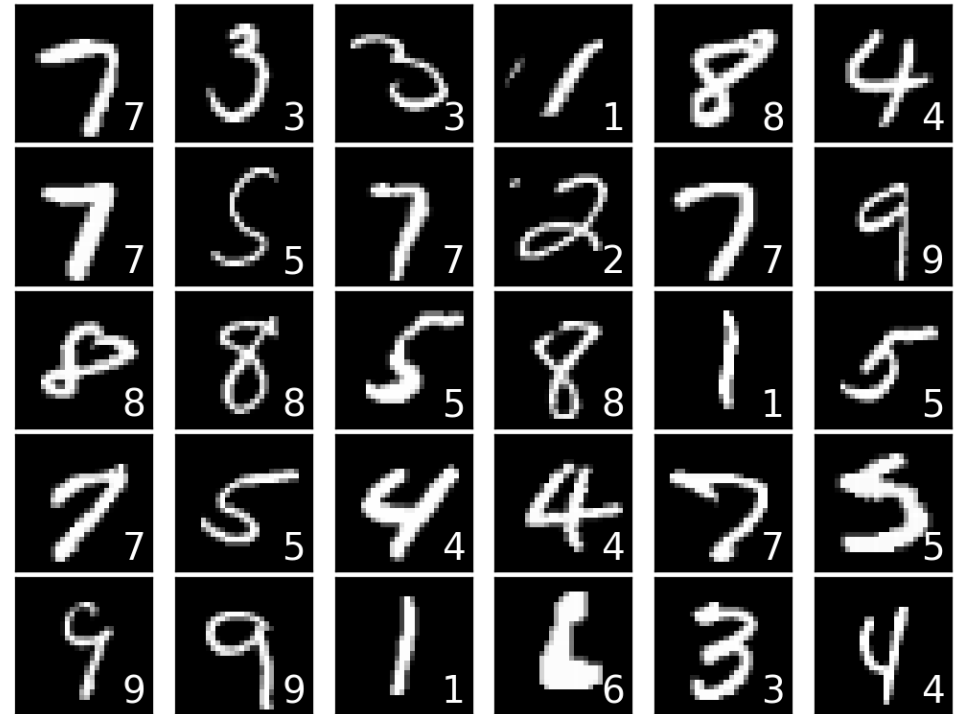
<https://projects.rajivshah.com/rldemo/>

MNIST Images (Pop Quiz)

• Task T : _____

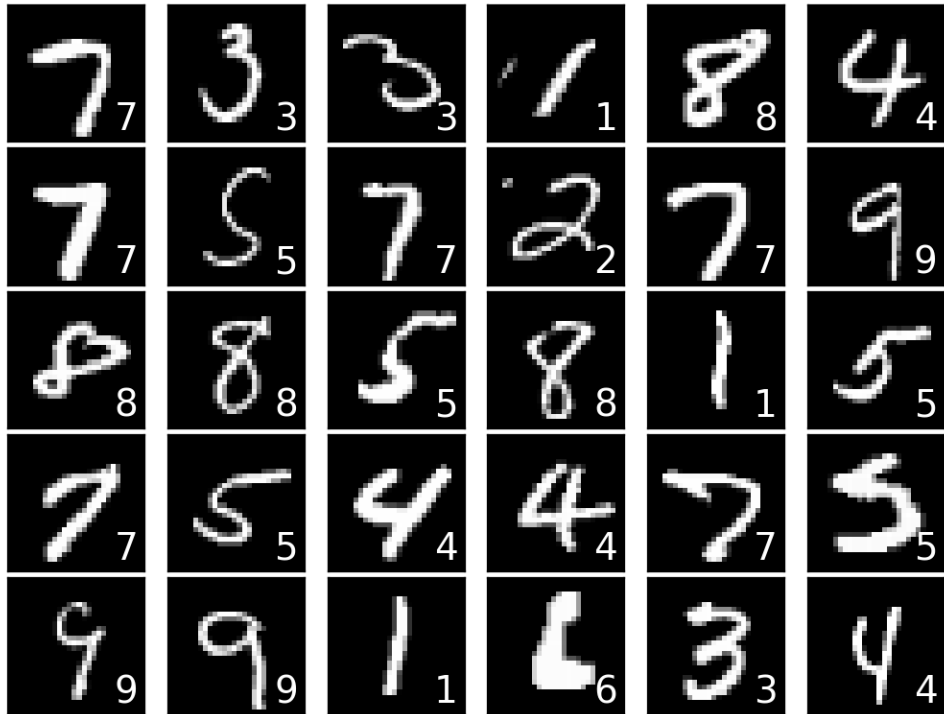
• Performance measure P : _____

• Training experience E : _____

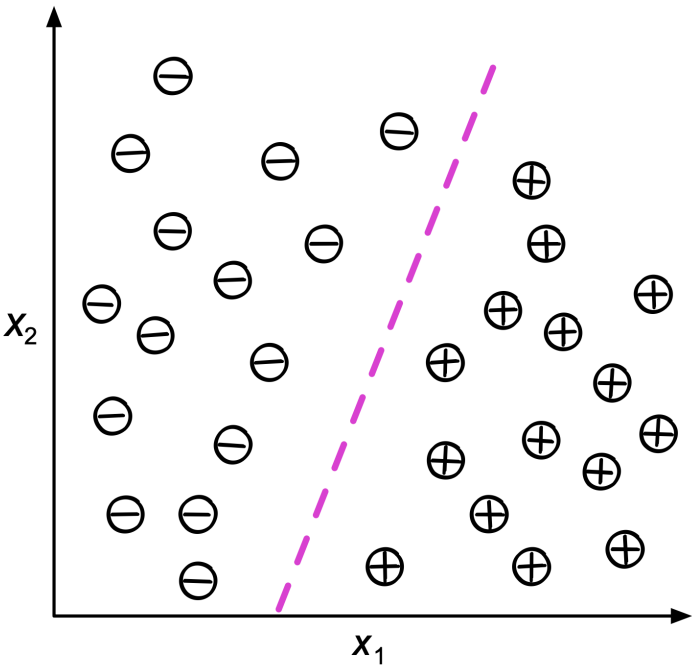


MNIST Images (Pop Quiz)

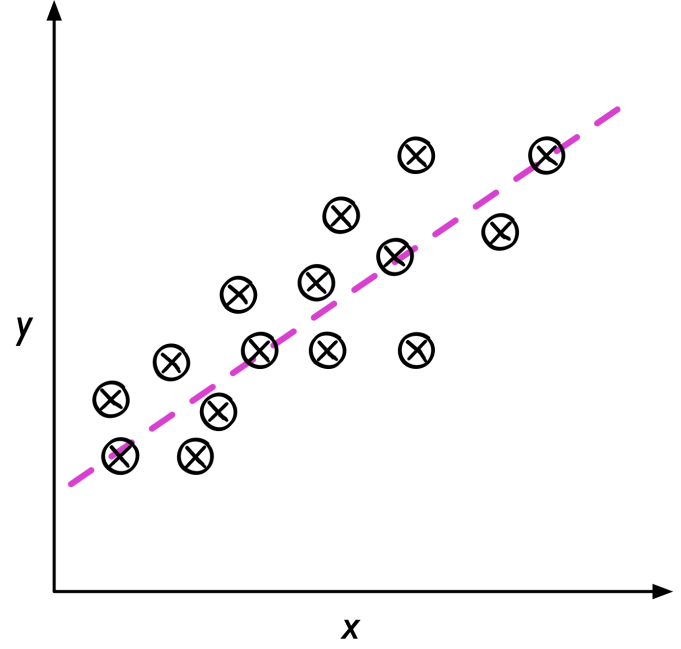
- Task T : Multi-class labeling
- Performance measure P : Accuracy, precision, etc.
- Training experience E : Edge histogram, correlation filter response, etc.



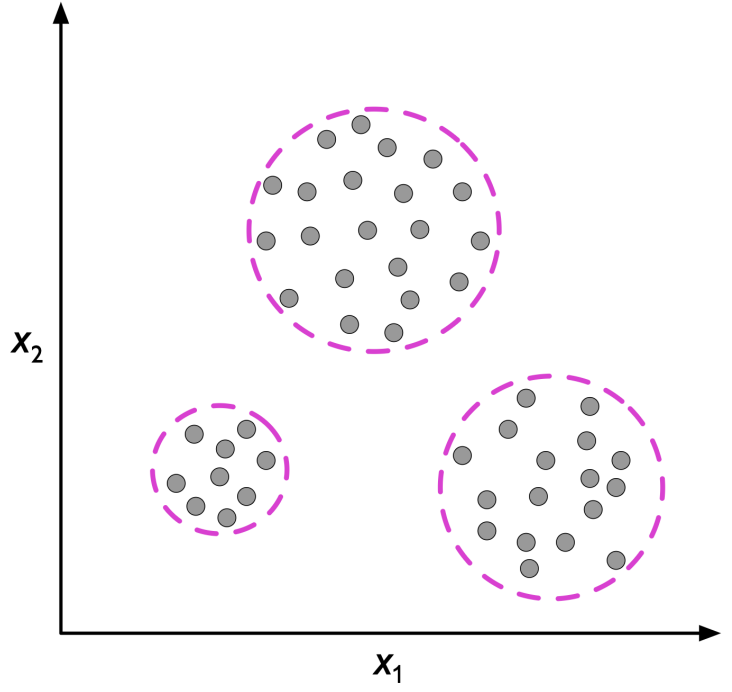
What do these plots have in common?



Classification



Regression



Clustering

Python

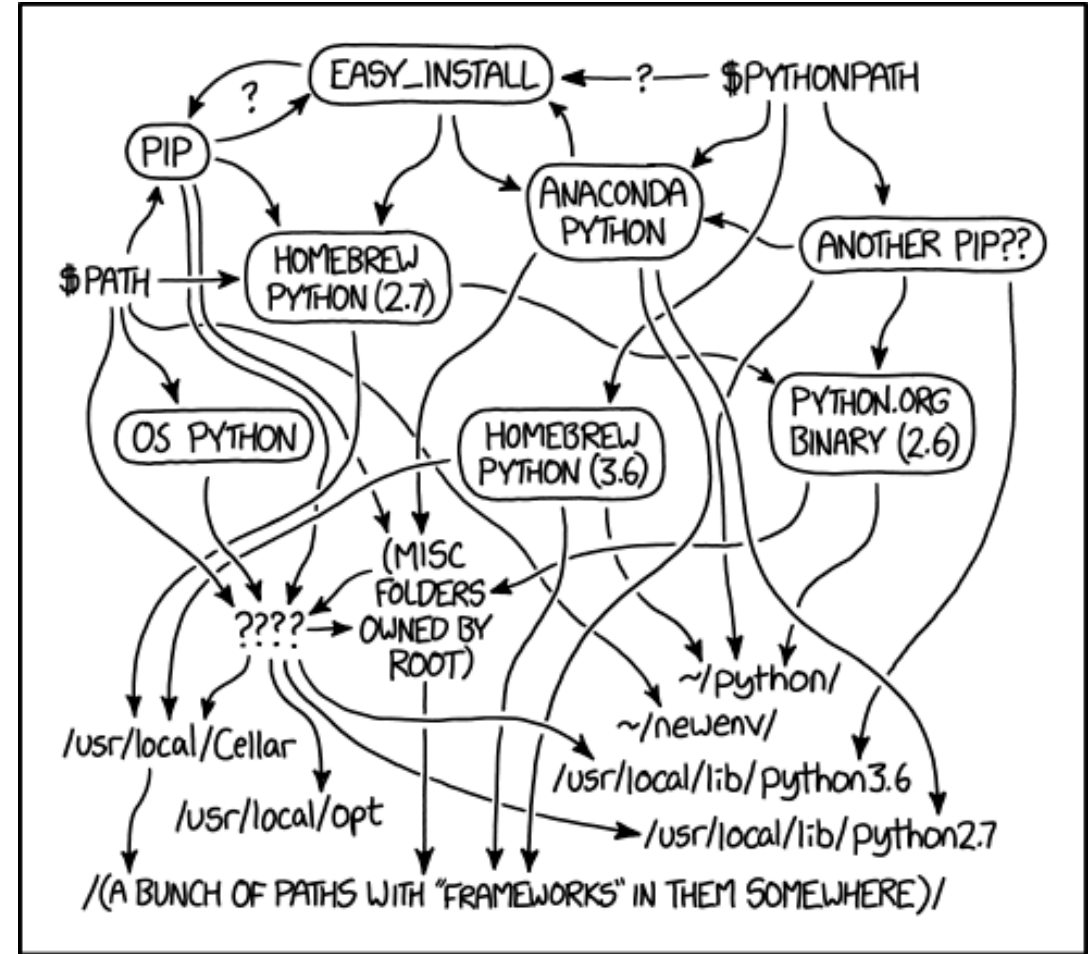
Why Python?

- **Extensive Libraries and Frameworks:** Rich ecosystem with tools like TensorFlow, PyTorch, and Scikit-Learn simplifies complex algorithm implementation.
- **Ease of Learning and Use:** Clear, readable syntax accelerates learning and allows rapid prototyping.
- **Active Community and Support:** Vast, active community offers extensive documentation, tutorials, and forums for support.
- **Versatility and Integration:** Seamlessly integrates with other languages and platforms, supporting various programming paradigms.
- **Data Handling Capabilities:** Efficient data structures and functions for handling large datasets through libraries like NumPy and Pandas.
- **Visualization Tools:** Powerful libraries like Matplotlib and Seaborn aid in data exploration and result presentation.
- **Industry Adoption and Support:** Widespread use in academia and industry ensures ongoing investment and support.



Setting up Python

- Python 3
- Environments:
 - Anaconda
 - PyEnv
 - Virtual Environments
- Pip
- Jupyter Notebooks
- Git



MY PYTHON ENVIRONMENT HAS BECOME SO DEGRADED THAT MY LAPTOP HAS BEEN DECLARED A SUPERFUND SITE.

Source: <https://xkcd.com/1987/>

Anaconda

- **Comprehensive Package Management:** Anaconda includes over 7,500 open-source pre-installed packages and simplifies the installation and management of additional packages through Conda, ensuring compatibility and reducing dependency conflicts.
- **Environment Management:** Easily create, manage, and switch between multiple isolated environments, allowing for clean and reproducible project setups without interfering with system-wide settings.
- **Integrated Development Environment:** Comes with Jupyter Notebook and Spyder, providing powerful, interactive tools for developing and testing code, data analysis, and visualization.
- **Cross-Platform Support:** Compatible with Windows, macOS, and Linux, ensuring consistent development environments across different operating systems.
- **Enhanced Data Science Tools:** Bundles essential libraries for machine learning, data analysis, and scientific computing (e.g., NumPy, Pandas, SciPy, and Matplotlib), streamlining the setup process for data science projects.



Important!

Don't mess with your operating system's Python installation.

Python Installation

- PyEnv in MacOS

```
# install pyenv requirements  
brew install openssl readline sqlite3 xz zlib
```

```
# install pyenv (add to PATH)  
brew install pyenv
```

```
# list available python versions  
pyenv versions
```

```
hsantosv@LAP513 ~ % pyenv versions  
  
* system (set by /Users/hsantosv/.pyenv/version)  
hsantosv@LAP513 ~ % where python3  
/opt/homebrew/bin/python3  
/usr/bin/python3
```


Python Installation

- PyEnv in MacOS

```
# install pyenv requirements  
brew install openssl readline sqlite3 xz zlib  
  
# install pyenv (add to PATH)  
brew install pyenv  
  
# list available python versions  
pyenv versions
```

Make sure you tell your OS where to find these tools.

```
hsantosv@LAP513 ~ % pyenv versions  
  
* system (set by /Users/hsantosv/.pyenv/version)  
hsantosv@LAP513 ~ % where python3  
/opt/homebrew/bin/python3  
/usr/bin/python3
```

Python Installation

- PyEnv in MacOS

```
# install python version 3.12.0  
pyenv install 3.12.0
```

```
# list available python versions  
pyenv versions
```

```
hsantosv@LAP513 ~ % pyenv versions  
  
* system (set by /Users/hsantosv/.pyenv/version)  
3.12.0
```

Python Installation

- PyEnv in MacOS

```
# install separate python version 3.12.5  
pyenv install 3.12.5
```

```
# set default version to newly installed 3.12.5  
pyenv global 3.12.5
```

```
# list available python versions  
pyenv versions
```

```
hsantosv@LAP513 ~ % pyenv versions  
  
system  
3.12.0  
* 3.12.5 (set by /Users/hsantosv/.pyenv/version)
```

Python Environment

- Good resource: <https://realpython.com/python-virtual-environments-a-primer/>
- Manage the libraries, python version, etc. for your project
- You can use conda, venv, or *pyenv-virtualenv*

```
# install pyenv-virtualenv plugin  
brew install pyenv-virtualenv
```

Python Environment

```
# create new virtualenv  
# (e.g. pyenv virtualenv <python-version> <env-name>)  
pyenv virtualenv 3.12.5 test_project
```

```
# activate the virtualenv  
pyenv activate test_project
```

```
# deactivate the virtualenv  
pyenv deactivate test_project
```

```
# list all available virtual environments  
pyenv virtualenvs
```

Python Environment

```
# create new virtualenv  
# (e.g. pyenv virtualenv <python-version> <env-name>)  
pyenv virtualenv 3.12.5 test_project
```

```
# activate the virtualenv  
pyenv activate test_project
```

```
# deactivate the virtualenv  
pyenv deactivate test_project
```

```
# list all available virtual environments  
pyenv virtualenvs
```

```
hsantosv@LAP513 .pyenv % pyenv virtualenvs  
3.12.0/envs/ml_clean_config (created from /Users/hsantosv/.pyenv/versions/3.12.0)  
3.12.5/envs/cosc325_ml_config (created from /Users/hsantosv/.pyenv/versions/3.12.5)  
3.12.5/envs/cosc525_dl_config (created from /Users/hsantosv/.pyenv/versions/3.12.5)  
3.12.5/envs/test_project (created from /Users/hsantosv/.pyenv/versions/3.12.5)  
cosc325_ml_config (created from /Users/hsantosv/.pyenv/versions/3.12.5)  
cosc525_dl_config (created from /Users/hsantosv/.pyenv/versions/3.12.5)  
ml_clean_config (created from /Users/hsantosv/.pyenv/versions/3.12.0)  
test_project (created from /Users/hsantosv/.pyenv/versions/3.12.5)
```


Local Python Environment

```
# Set a directorate environment to specific Python Environment  
cd local_directory  
pyenv local cosc325_ml_config
```

```
hsantosv@LAP513 Development %  
hsantosv@LAP513 Development % cd /Users/hsantosv/Library/CloudStorage/OneDrive-UniversityofTennessee/04-Teaching/UTK-TCH-COSC-325/Development  
hsantosv@LAP513 Development % pyenv local cosc325_ml_config  
[(cosc325_ml_config) hsantosv@LAP513 Development % cd ,,
```

PIP COSC325 Libraries

install jupyterlab into your virtual environment

```
pip install jupyterlab
```

install Numpy for, Matlab-like, matrix multiplication capabilities

```
pip install numpy
```

```
pip install scipy          #optional: expands numpy functions
```

install Pandas for manipulation of tabulated data and seaborn for pandas-supported plots

```
pip install pandas
```

```
pip install seaborn
```

install Scikit-Learn for most classical machine learning needs

```
pip install scikit-learn
```

install matplotlib for general generation of graphs/plots

```
pip install matplotlib
```

install PyTensor (formaly Theano) to optimize matrix computations (e.g., use of GPUs)

```
pip install pytensor
```

Saving/Exporting Your Environment

```
# save environment libraries to file  
pip freeze > requirements.txt
```

```
# EXPORT  
# Activate environment  
pyenv activate <env_name>
```

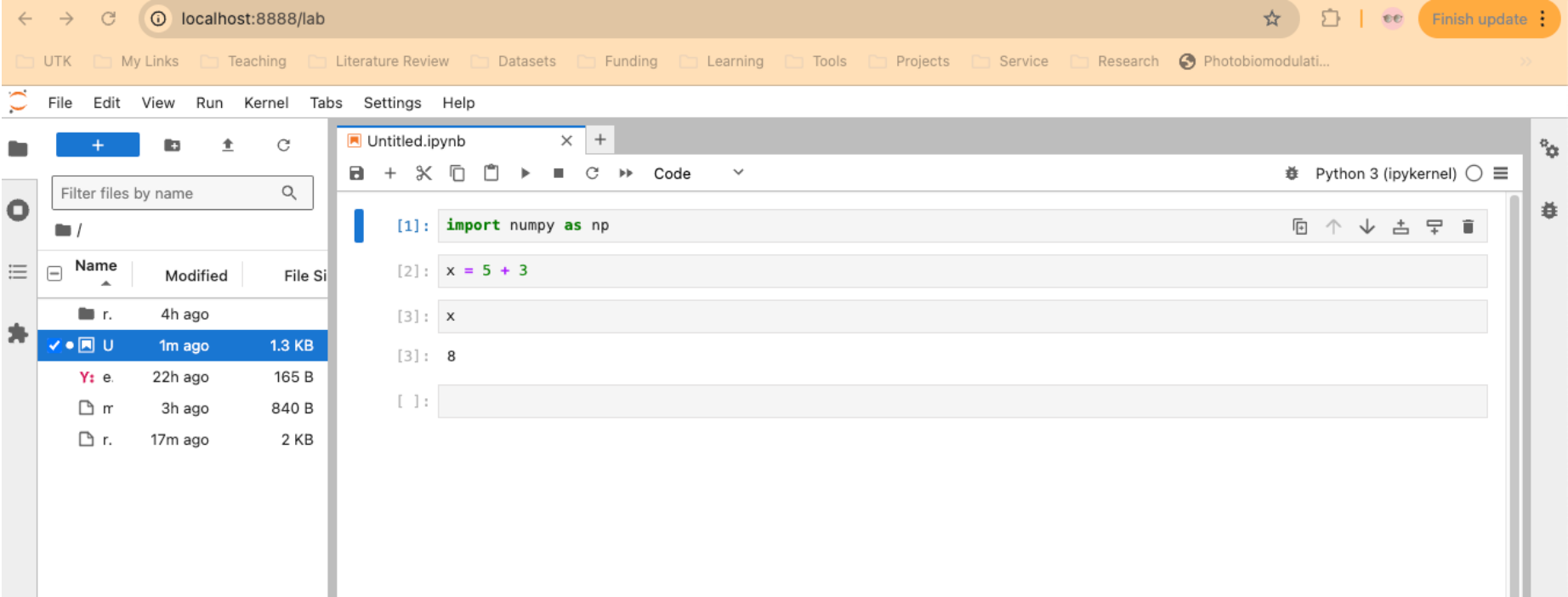
```
# import environment libraries from file  
(<env_name>)$ pip install -r path/to/requirements.txt
```

Note that it does not save the Python version.

```
anyio==4.4.0  
appnope==0.1.4  
argon2-cffi==23.1.0  
argon2-cffi-bindings==21.2.0  
arrow==1.3.0  
asttokens==2.4.1  
async-lru==2.0.4  
attrs==24.2.0  
babel==2.16.0  
beautifulsoup4==4.12.3  
bleach==6.1.0  
certifi==2024.7.4  
cffi==1.17.0  
charset-normalizer==3.3.2  
comm==0.2.2  
cons==0.4.6  
contourpy==1.2.1  
cyclery==0.12.1  
debugpy==1.8.5  
decorator==5.1.1  
defusedxml==0.7.1  
etuples==0.3.9  
executing==2.0.1  
fastjsonschema==2.20.0  
filelock==3.15.4  
fonttools==4.53.1  
fqdn==1.5.1  
h11==0.14.0  
httpcore==1.0.5  
httpx==0.27.0  
idna==3.7  
ipykernel==6.29.5  
ipython==8.26.0  
isoduration==20.11.0  
jedi==0.19.1  
Jinja2==3.1.4  
joblib==1.4.2  
json5==0.9.25  
jsonpointer==3.0.0  
jsonschema==4.23.0  
jsonschema-specifications==2023.12.1  
jupyter-events==0.10.0  
jupyter-lsp==2.2.5  
jupyter_client==8.6.2  
jupyter_core==5.7.2  
jupyter_server==2.14.2  
jupyter_server_terminals==0.5.3  
jupyterlab==4.2.4  
jupyterlab_pygments==0.3.0  
jupyterlab_server==2.27.3  
kiwisolver==1.4.5  
logical-unification==0.4.6  
MarkupSafe==2.1.5  
matplotlib==3.9.2  
matplotlib-inline==0.1.7  
miniKanren==1.0.3  
mistune==3.0.2  
multipledispatch==1.0.0  
nbclient==0.10.0  
nbconvert==7.16.4  
nbformat==5.10.4  
nest-asyncio==1.6.0  
notebook_shim==0.2.4  
numpy==1.26.4  
overrides==7.7.0  
packaging==24.1  
pandas==2.2.2  
pandocfilters==1.5.1  
parso==0.8.4  
pexpect==4.9.0  
pillow==10.4.0  
platformdirs==4.2.2  
prometheus_client==0.20.0  
prompt_toolkit==3.0.47  
psutil==6.0.0  
ptyprocess==0.7.0  
pure_eval==0.2.3  
pyparsing==2.2.2  
Pygments==2.18.0
```

Jupyter Notebook

```
# Open jupyter server  
jupyter lab
```



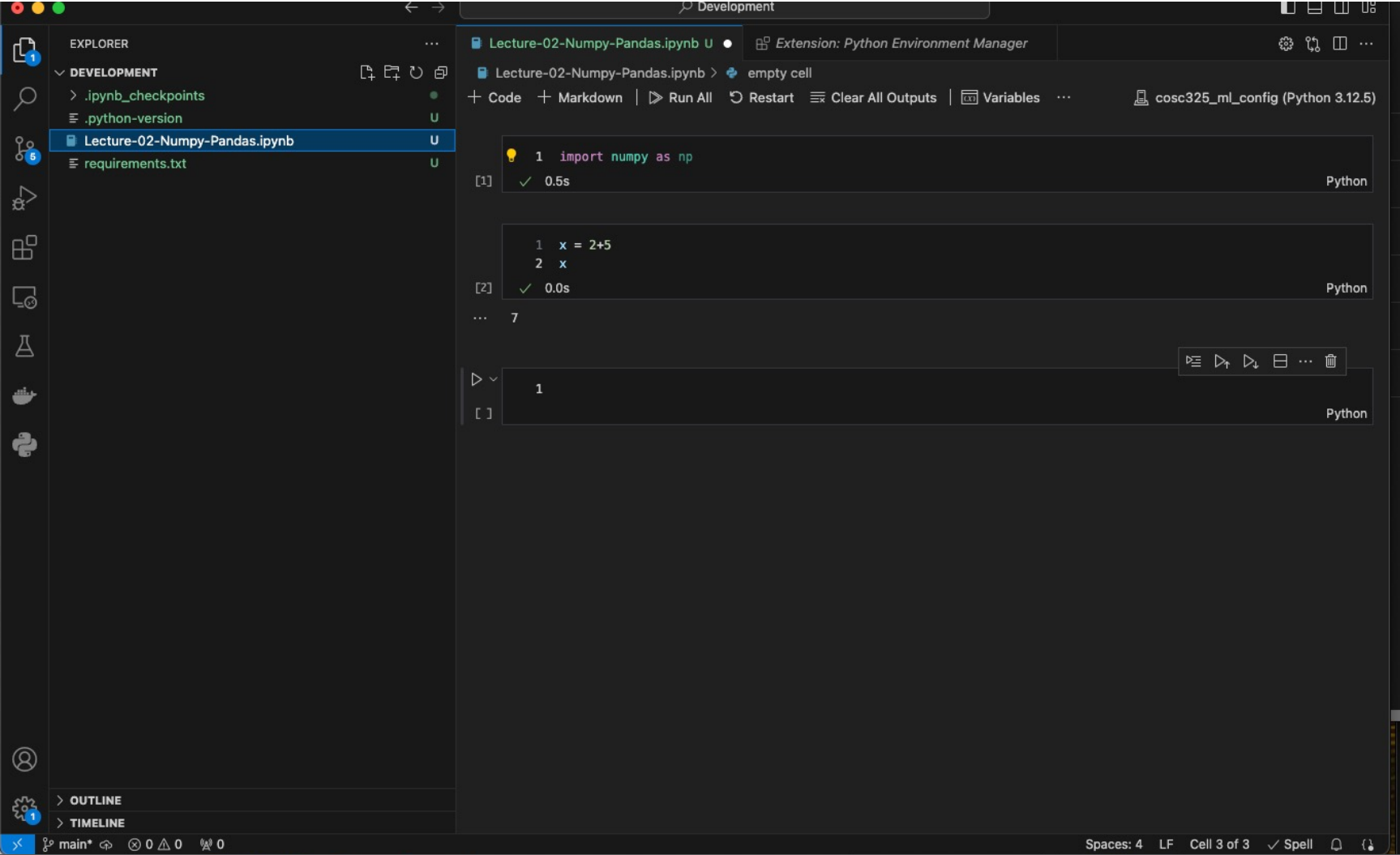
Jupyter Notebooks

- **Interactive Development:** Jupyter Notebooks allow real-time code execution and visualization, enabling immediate feedback and iterative development, which is ideal for data exploration and analysis.
- **Rich Media Support:** Integrates code, text, images, and visualizations in a single document, enhancing documentation and presentation of data analysis and research findings.
- **Collaborative Features:** Facilitates collaboration by enabling easy sharing and version control of notebooks, allowing multiple users to work on the same document seamlessly.
- **Wide Language Support:** Supports over 40 programming languages, including Python, R, and Julia, making it versatile for various data science and computational tasks.

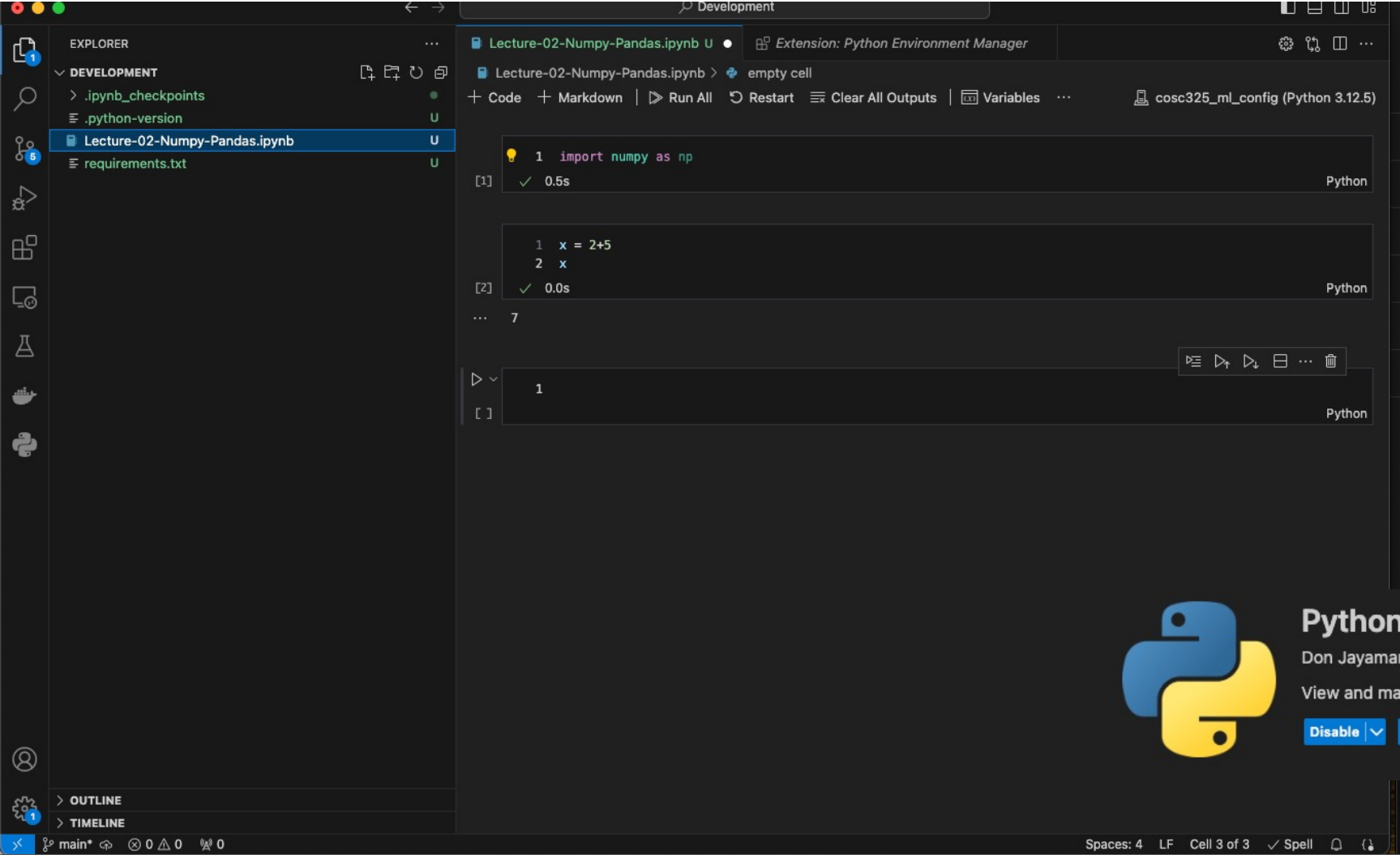


```
pip install jupyterlab
```

VSCode



VSCode



Inside VSCode:

- CMD+Shift+P
- Type: shell command
- Select: Install 'code' command in Path

Other IDEs

- PyCharm [Local]
 - <https://www.jetbrains.com/pycharm/>
- Google Colab
 - <https://colab.research.google.com/>
- Anaconda Notebooks
 - <https://anaconda.cloud/code-in-the-cloud>
- Kaggle
 - <https://www.kaggle.com/code>
- JupyterLab
 - <https://jupyter.org/try-jupyter/lab/>

Python Basics

- General and lambda functions
 - <https://www.geeksforgeeks.org/python-functions/>
- Classes
 - <https://www.geeksforgeeks.org/python-classes-and-objects/>
- Loops and how to iterate over lists
 - <https://www.geeksforgeeks.org/loops-in-python/>
- Data types, such as Lists, Tuples, Dictionaries
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 - <https://www.geeksforgeeks.org/python-tuples/>
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 - <https://www.geeksforgeeks.org/python-dictionary/>

<https://www.geeksforgeeks.org/python-difference-between-list-and-tuple/>

Lecture Recap

- Machine learning
 - A subfield of artificial intelligence
 - Models need to generalize (i.e., learn)
 - Task, Experience, Performance
 - Different learning categories: supervised, self-supervised, and reinforcement.
- Programming
 - Python: flexible, efficient, collaborative, powerful
 - Always work from a dockerized or virtual environment
 - Practice

Next Week

- Scientific computing with Python
 - Deep dive for Numpy, Pandas, and Scikit-Learn libraries
- Bring your laptops with Python environment ready.
 - Check tutorial in Canvas