

Introduction to Deep Learning

With PyTorch

What is AI?

What is Artificial Intelligence?

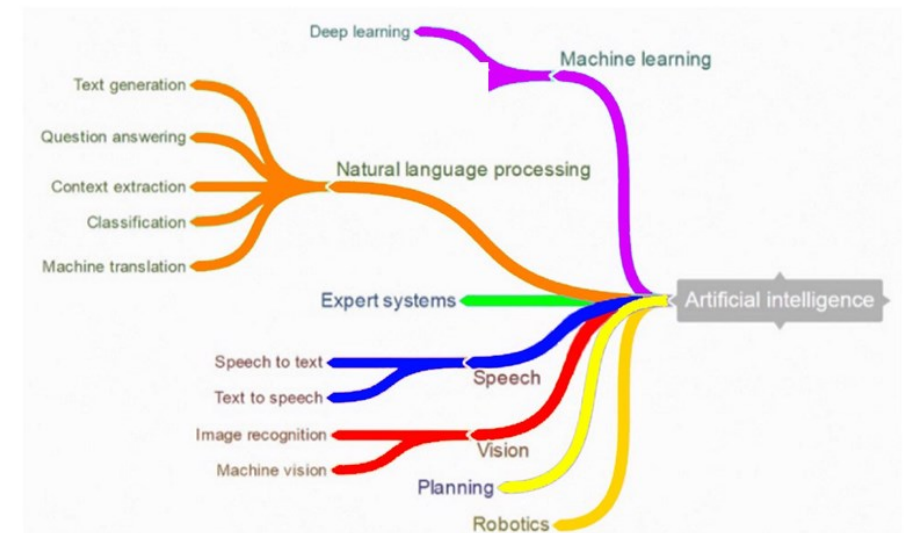
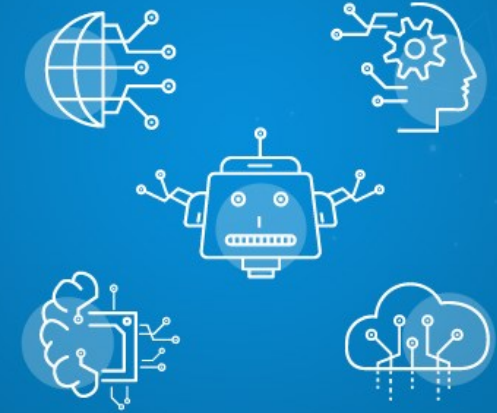
Artificial Intelligence is the field of computer science that is associated with the concept of machines “thinking like humans” to perform tasks such as learning, problem-solving, planning, reasoning and identifying patterns.

Customer Insight Group, Inc.

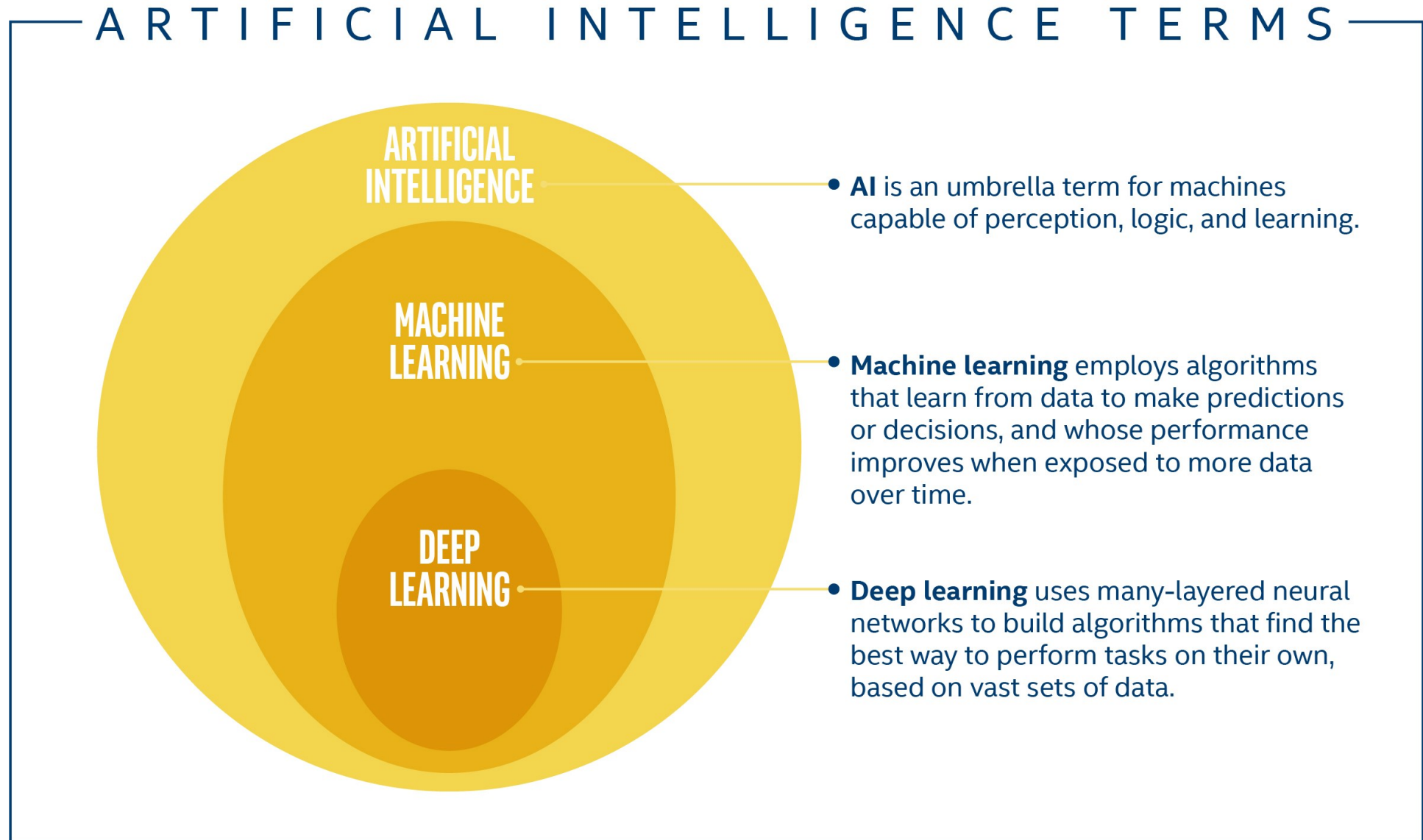
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Types Of Artificial Intelligence

edureka!

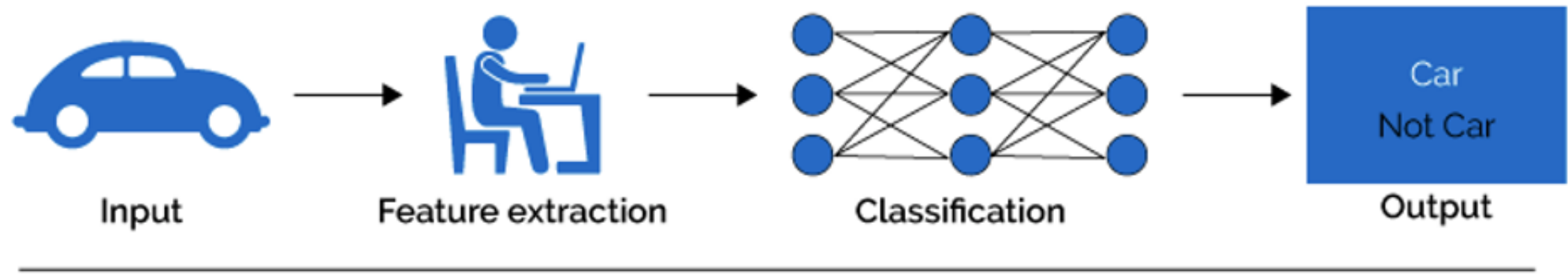


AI, ML, DL?

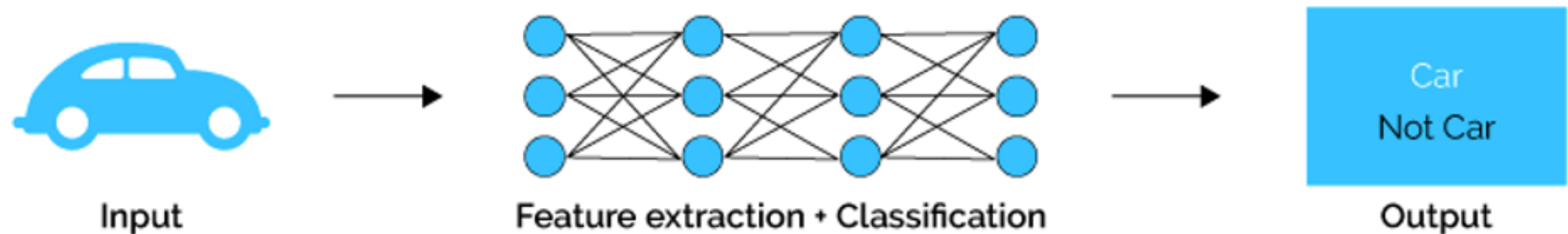


ML and DL?

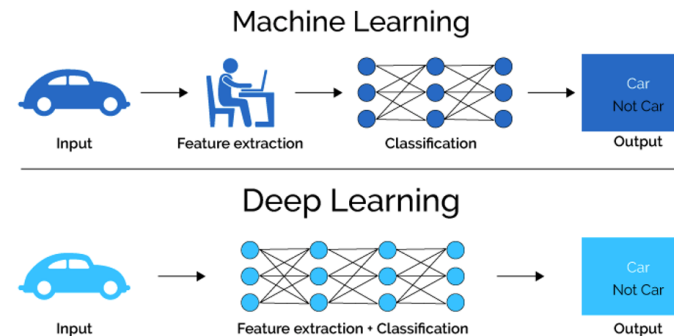
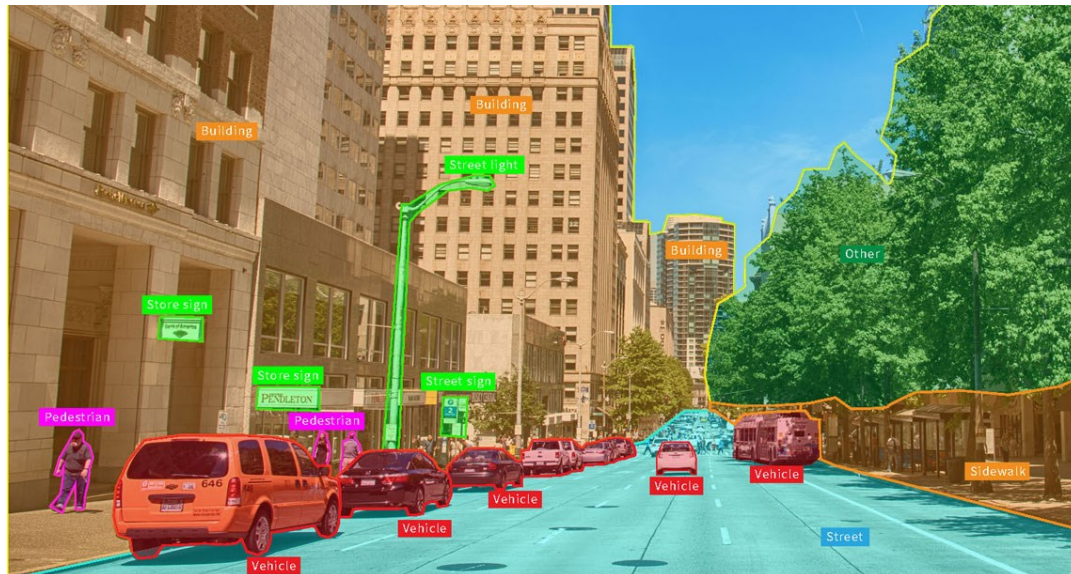
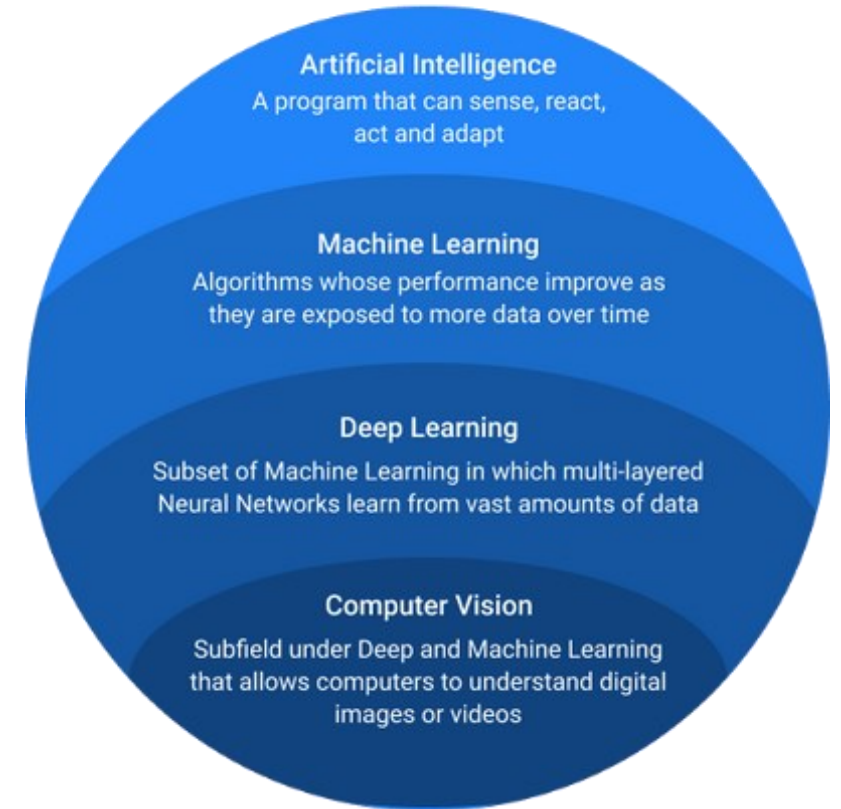
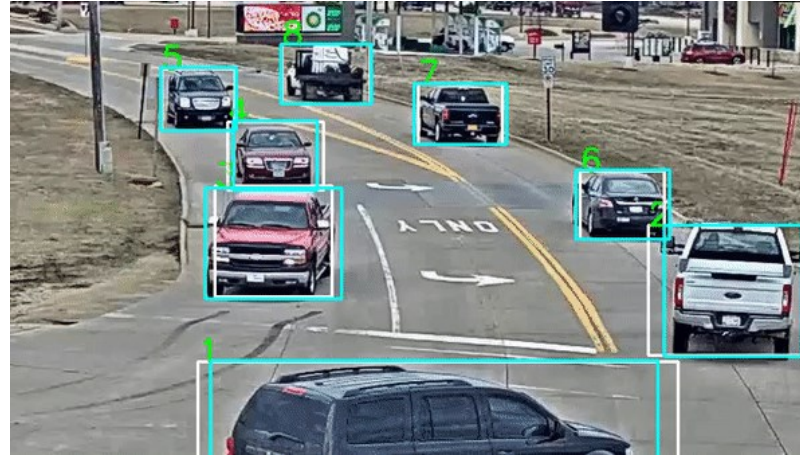
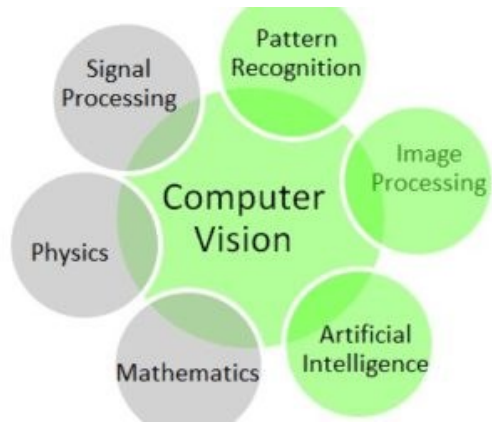
Machine Learning



Deep Learning



DL and CV?



Tasks Related to DL/ CV

- Classification
- Regression
- Detection
- Tracking
- Segmentation
- Registration
- 3D Reconstruction
 - 3D Computer Vision

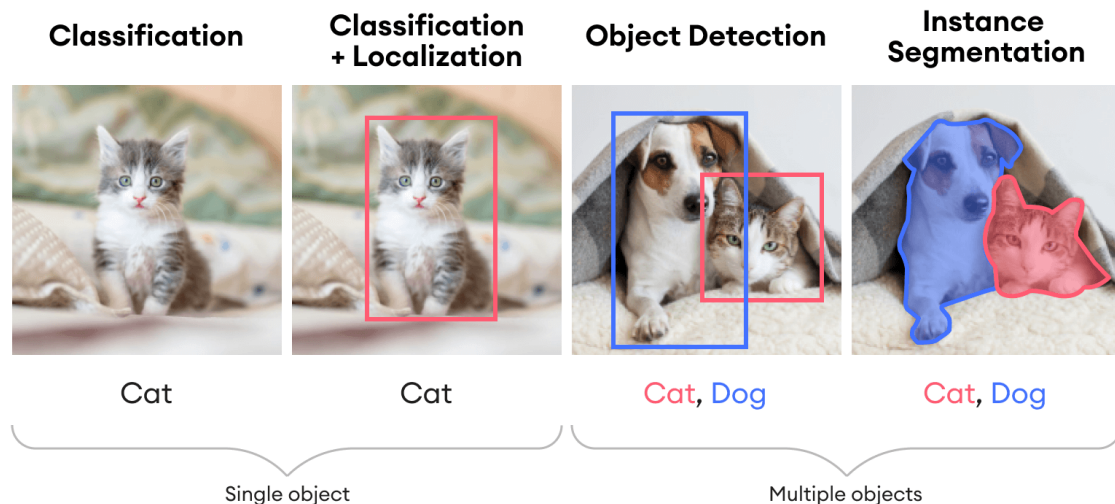
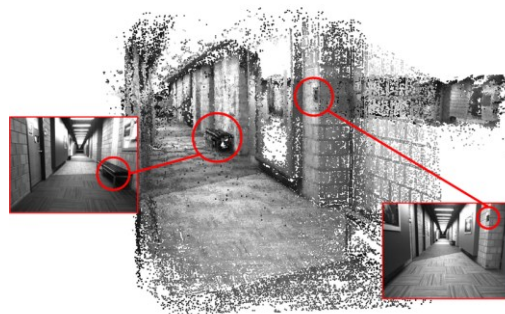
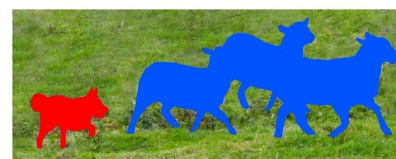
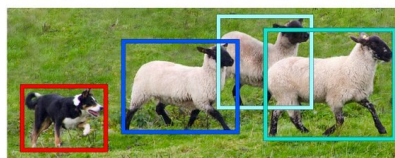


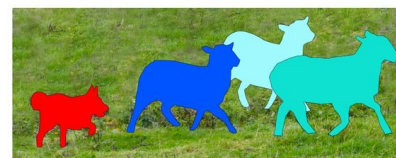
Image Recognition



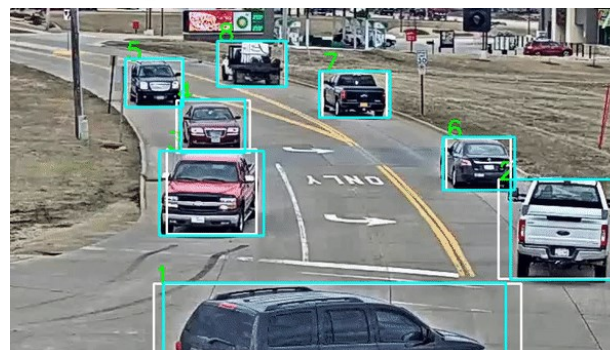
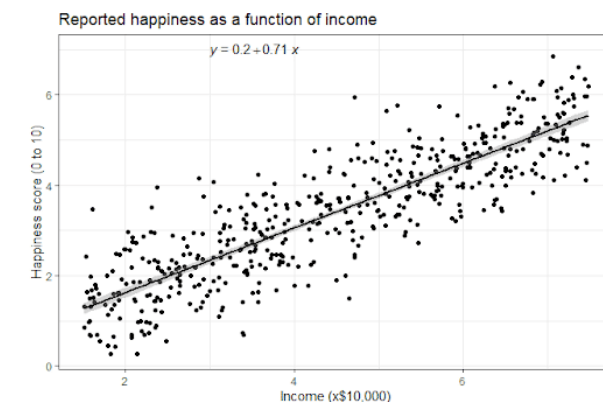
Semantic Segmentation



Object Detection

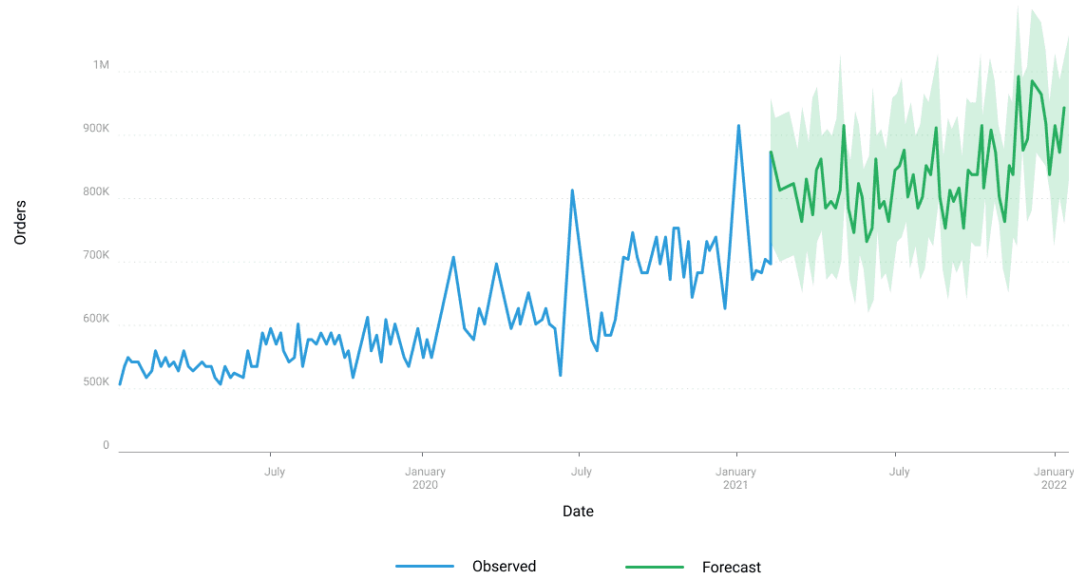


Instance Segmentation



Is DL a Remedy for Everything?

- Decision-making
- Forecasting
- Predictions/ Uncertainty Analysis

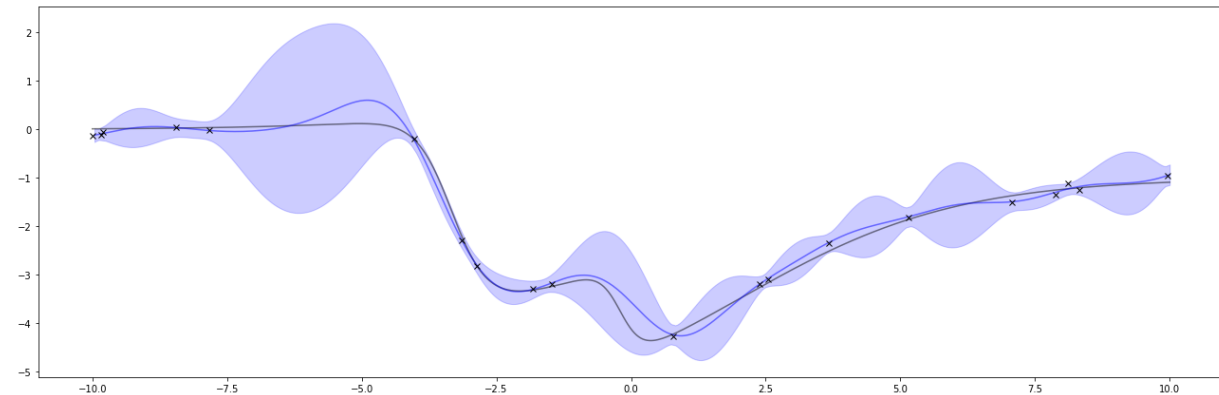


mob/dev

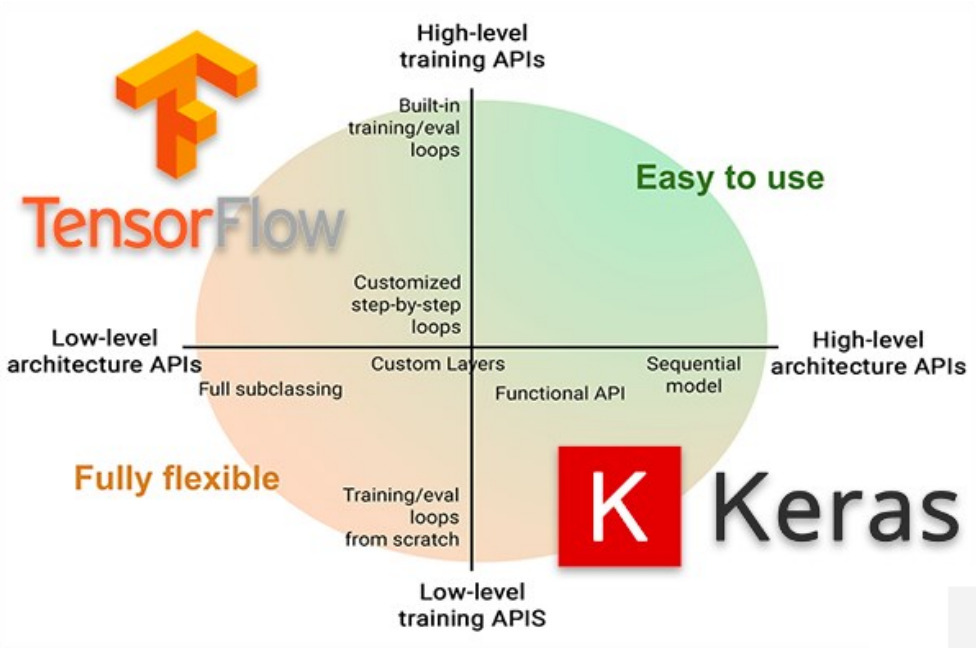
Condition
Any expression that evaluates to true or false

```
if (condition) {  
    statement  
    statement  
    ...  
}  
following_statement
```

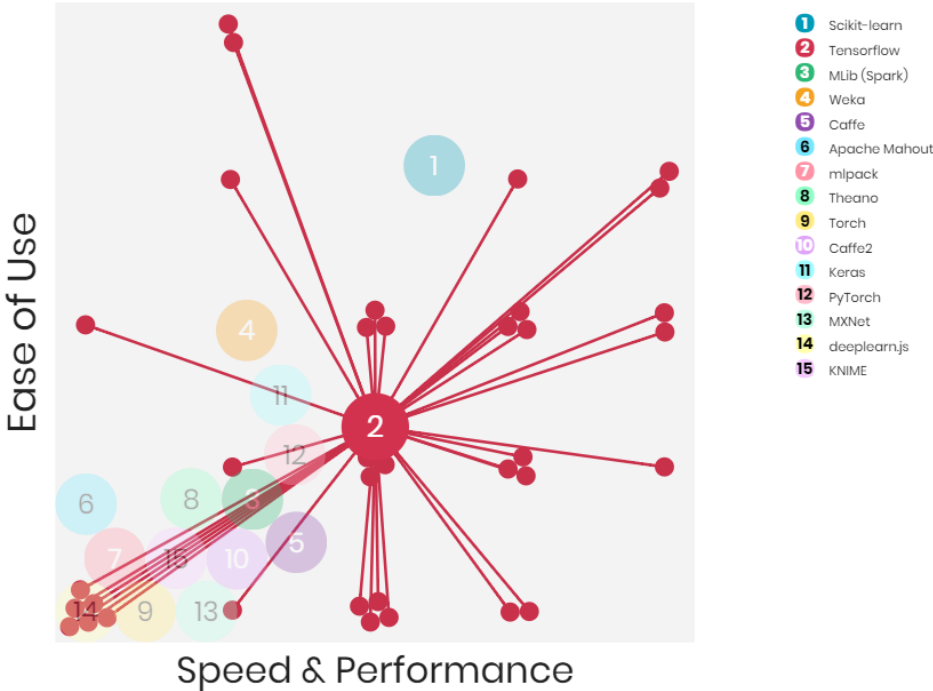
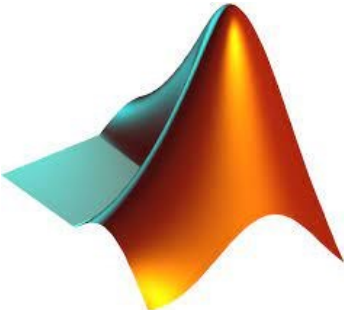
True branch
This is executed if the condition is true



DL Frameworks



TensorFlow



How To Start Solving DL Problems?

1. Get a Good GPU and RAM for model training (or use cloud)
2. Install OS, NVIDIA drivers, CUDA, cuDNN libs
3. Install Anaconda/ Python. Create Environment and Select DL Framework
 - OS: Windows/ Ubuntu (Recommended)
4. Setup development environment
 - Install libraries and their dependencies
 - Pay attention to versions
5. Understand the problem/ Prepare or download the datasets related to the task
 - For custom datasets, first acquire it and then annotate it
 - Annotation Tools: MATLAB ImageLabeler, LabelMe etc.
6. Model Training/ Testing
 - Model Evaluation

Anaconda

ANACONDA.NAVIGATOR

Connect

Home

Environments

Learning


Community

All applications

on

base (root)


Channels



DataSpell

DataSpell is an IDE for exploratory data analysis and prototyping machine learning models. It combines the interactivity of Jupyter notebooks with the intelligent Python and R coding assistance of PyCharm in one user-friendly environment.


Install



DataLore

Online Data Analysis Tool with smart coding assistance by JetBrains. Edit and run your Python notebooks in the cloud and share them with your team.


Launch



IBM Watson Studio Cloud

IBM Watson Studio Cloud provides you the tools to analyze and visualize data, to cleanse and shape data, to create and train machine learning models. Prepare data and build models, using open source data science tools or visual modeling.

Launch




JupyterLab

3.3.2

An extensible environment for interactive and reproducible computing, based on the Jupyter Notebook and Architecture.

Launch




Jupyter Notebook

6.4.8

Web-based, interactive computing notebook environment. Edit and run human-readable docs while describing the data analysis.

Launch




PyCharm Community

2022.1.1

An IDE by JetBrains for pure Python development. Supports code completion, listing, and debugging.

Launch




Qt Console

5.3.0

PyQt GUI that supports inline figures, proper multiline editing with syntax highlighting, graphical calltips, and more.

Launch




Spyder

5.1.5

Scientific Python Development Environment. Powerful Python IDE with advanced editing, interactive testing, debugging and introspection features

Launch




VS Code

1.68.1

Streamlined code editor with support development operations like debugging, task running and version control.


Launch



Google Colab

Cloud-based Jupyter environment with free GPU access.


Launch



Orange3

Data mining toolbox with visual interfaces and Python scripts.


Launch



RStudio

R IDE for statistical computing and graphics.

Launch



RStudio

1.1.456

A set of integrated tools designed to help you be more productive with R. Includes R essentials and notebooks.

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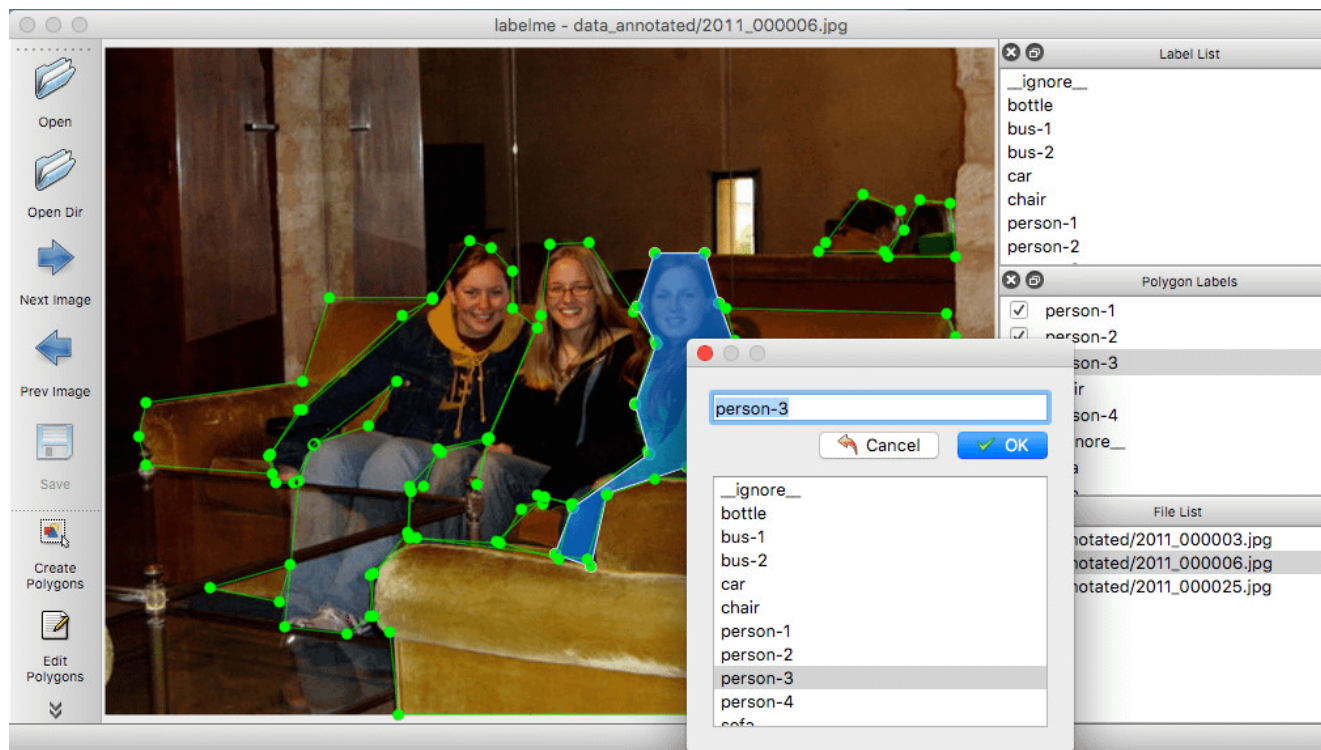
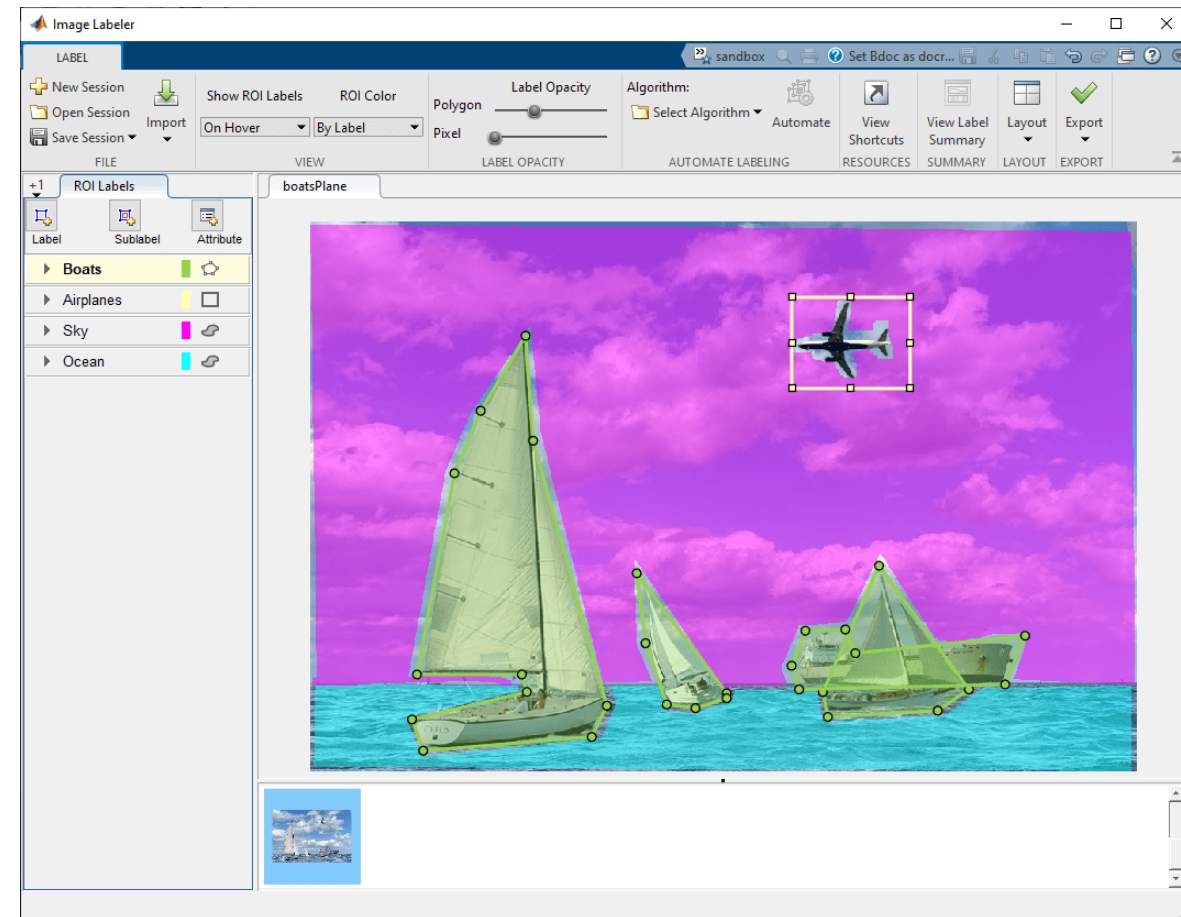
Anaconda Prompt (Anaconda3) - conda activate base

```
-----
train Loss: 0.3456 Acc: 0.8701
test Loss: 1.2231 Acc: 0.6570
Epoch 3/4
-----
train Loss: 0.2178 Acc: 0.9350
test Loss: 1.0505 Acc: 0.6950
Epoch 4/4
-----
train Loss: 0.2262 Acc: 0.9272
test Loss: 0.9066 Acc: 0.7750
Training complete in 4m 35s
Best test Acc: 0.935039
Traceback (most recent call last):
  File "C:\DLWorkshop\TrainWithCustomData.py", line 228, in <module>
    main()
NameError: name 'main' is not defined. Did you mean: 'min'?

(workshop) C:\DLWorkshop>conda activate base

(base) C:\DLWorkshop>
```

Ground Truth Annotation Tools



Example: A Simple Classification Network

Step # 1: Load Dataset

```
import torch
import torchvision
import torchvision.transforms as transforms

transform = transforms.Compose(
    [transforms.ToTensor(),
     transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))])

batch_size = 4

trainset = torchvision.datasets.CIFAR10(root='./data', train=True,
                                         download=True, transform=transform)
trainloader = torch.utils.data.DataLoader(trainset, batch_size=batch_size,
                                           shuffle=True, num_workers=2)

testset = torchvision.datasets.CIFAR10(root='./data', train=False,
                                         download=True, transform=transform)
testloader = torch.utils.data.DataLoader(testset, batch_size=batch_size,
                                          shuffle=False, num_workers=2)

classes = ('plane', 'car', 'bird', 'cat',
           'deer', 'dog', 'frog', 'horse', 'ship', 'truck')
```

Example: A Simple Classification Network

Step # 2: Create the Network and Optimizer

```
import torch.optim as optim
```

```
criterion = nn.CrossEntropyLoss()
```

```
optimizer = optim.SGD(net.parameters(), lr=0.001, momentum=0.9)
```

```
import torch.nn as nn
import torch.nn.functional as F
```

```
class Net(nn.Module):
    def __init__(self):
        super().__init__()
        self.conv1 = nn.Conv2d(3, 6, 5)
        self.pool = nn.MaxPool2d(2, 2)
        self.conv2 = nn.Conv2d(6, 16, 5)
        self.fc1 = nn.Linear(16 * 5 * 5, 120)
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)

    def forward(self, x):
        x = self.pool(F.relu(self.conv1(x)))
        x = self.pool(F.relu(self.conv2(x)))
        x = torch.flatten(x, 1) # flatten all dimensions except batch
        x = F.relu(self.fc1(x))
        x = F.relu(self.fc2(x))
        x = self.fc3(x)
        return x
```

```
net = Net()
```

Example: A Simple Classification Network

Step # 3: Write Training Routine

```
PATH = './cifar_net.pth'  
torch.save(net.state_dict(), PATH)
```

```
for epoch in range(2): # loop over the dataset multiple times  
  
    running_loss = 0.0  
    for i, data in enumerate(trainloader, 0):  
        # get the inputs; data is a list of [inputs, labels]  
        inputs, labels = data  
  
        # zero the parameter gradients  
        optimizer.zero_grad()  
  
        # forward + backward + optimize  
        outputs = net(inputs)  
        loss = criterion(outputs, labels)  
        loss.backward()  
        optimizer.step()  
  
        # print statistics  
        running_loss += loss.item()  
        if i % 2000 == 1999: # print every 2000 mini-batches  
            print(f'[{epoch + 1}, {i + 1:5d}] loss: {running_loss / 2000:.3f}')  
            running_loss = 0.0  
  
    print('Finished Training')
```

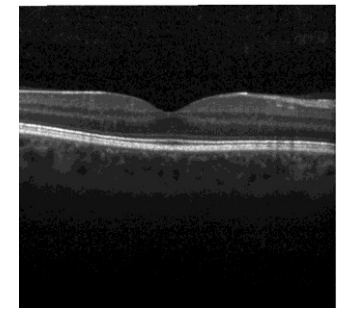
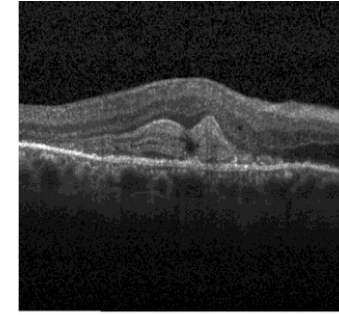

Example: A Simple Classification Network

Step # 4: Test and Evaluate the Trained Model

```
net = Net()  
net.load_state_dict(torch.load(PATH))
```

```
correct_pred = {classname: 0 for classname in classes}  
total_pred = {classname: 0 for classname in classes}  
  
# again no gradients needed  
with torch.no_grad():  
    for data in testloader:  
        images, labels = data  
        outputs = net(images)  
        _, predictions = torch.max(outputs, 1)  
        # collect the correct predictions for each class  
        for label, prediction in zip(labels, predictions):  
            if label == prediction:  
                correct_pred[classes[label]] += 1  
            total_pred[classes[label]] += 1  
  
# print accuracy for each class  
for classname, correct_count in correct_pred.items():  
    accuracy = 100 * float(correct_count) / total_pred[classname]  
    print(f'Accuracy for class: {classname:5s} is {accuracy:.1f} %')
```

Exercise (Image Classification): Training Model on Custom Dataset



- Use the given dataset to predict retinal diseases.
 - [Link](#)

Steps:

- Install OS, Anaconda, Python
- Load Anaconda and Create Environment
- Install Pip, PyTorch, Torchvision, pandas, scikit-image, matplotlib and all the dependencies
- Download the code file 'TrainWithCustomData.py' from the GitHub
- Run it and compute the results (in terms of accuracy)
- Understand the code and explain it in the next session
- Use to same code to classify digits in MNIST dataset