



PLURALSIGHT

Intro to AI/ML in Azure

Welcome!



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Objectives

At the end of this course, you will be able to:

- Have a good understanding of Azure **Machine Learning Workspace** and **Azure AI Services**
- Be ready to continue your journey and explore taking **DP-100** and **A1-102** Certification



Microsoft Certified: Azure AI Engineer Associate



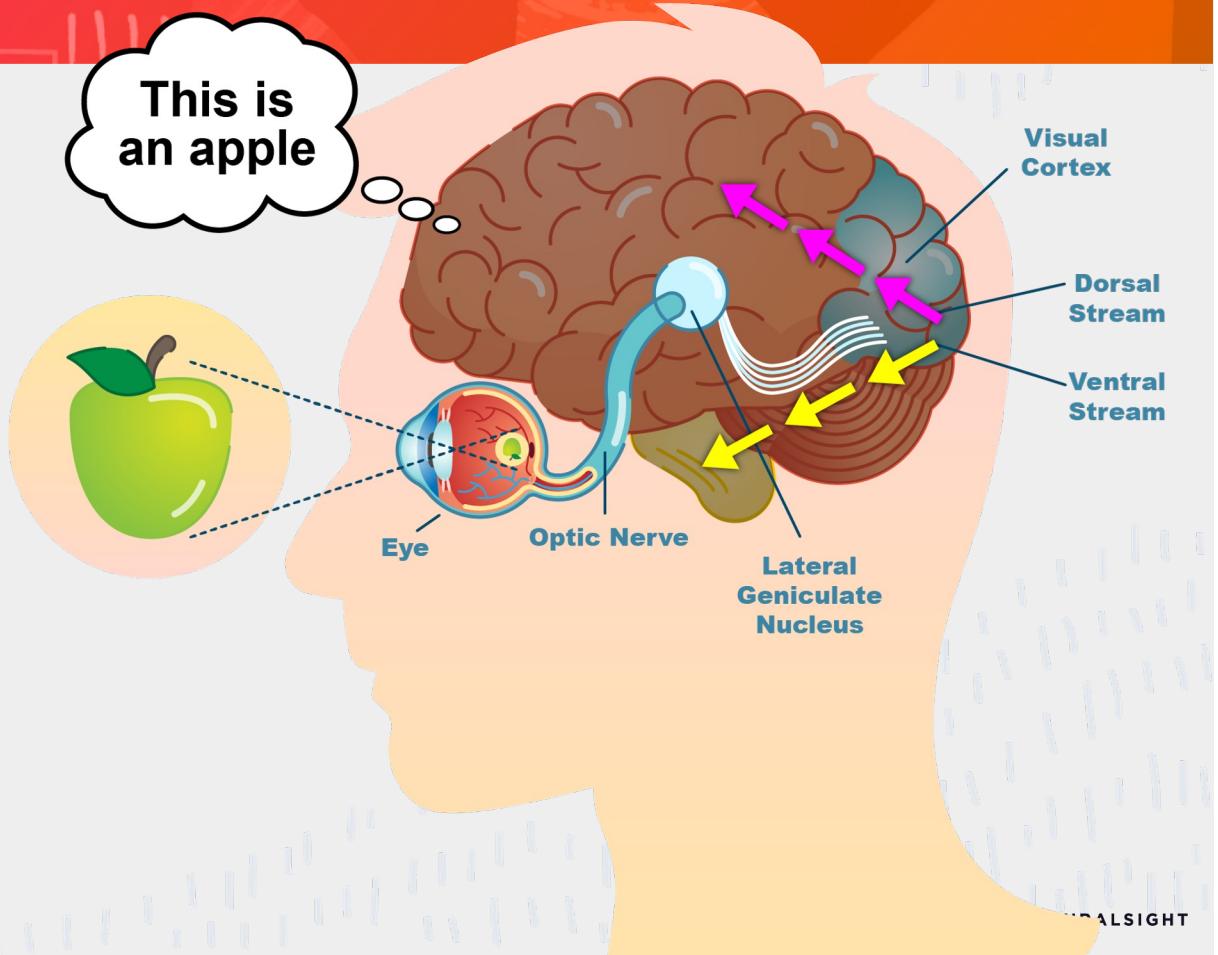
Microsoft Certified: Azure Data Scientist Associate

Artificial Neural Networks (ANN)

Neural Networks

How our brain works:

In order to recognize an image, our brain uses thousands of neuron connections to find a match between the visual input and a mental representation of an object.

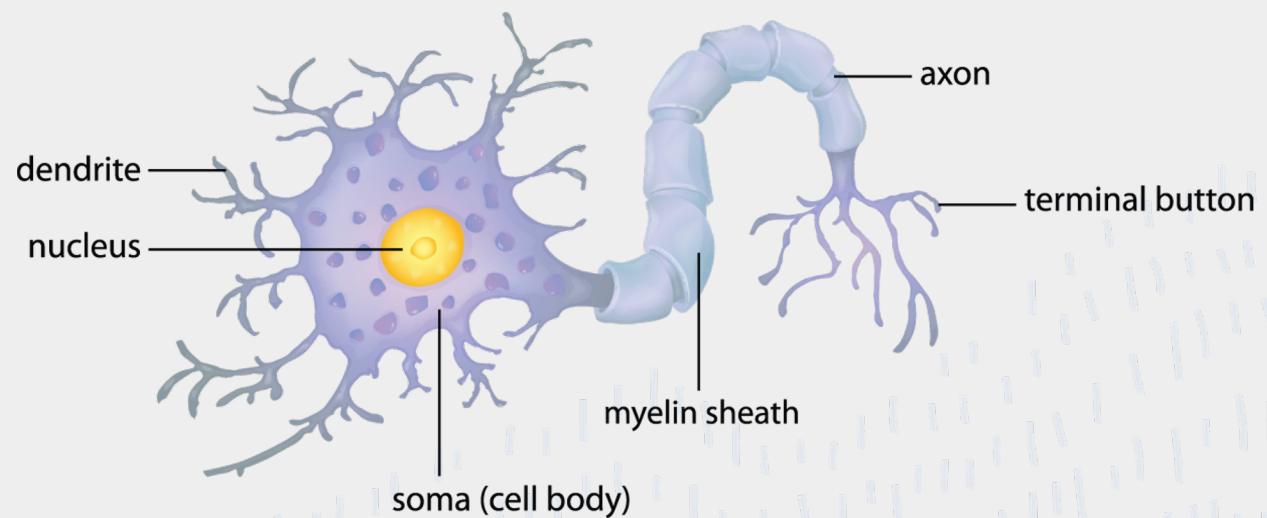


Neural Networks

The ability of the brain to process information and make predictions or interpretations is what inspired neurophysiologists and mathematicians to start the development of artificial neural networks (ANN).

In the same way that biological neurons receive input signals through the dendrites, an ANN receives input variables and processes them by using an activation function.

The output of an ANN is similar to the neuron nucleus in the brain.



History of Neural Networks

1943

Neurophysiologist **Warren McCulloch** and mathematician **Walter Pitts** wrote a paper on how neurons might work.

1949

Donald Hebb wrote *The Organization of Behavior*, which pointed out the fact that neural pathways are strengthened each time they are used.

1959

Bernard Widrow and **Marcian Hoff** of Stanford developed models called ADALINE and MADALINE.

1962

Widrow and **Hoff** developed a learning procedure that examines the value before the weight adjusts it (i.e., 0 or 1) according to the rule: Weight Change = (Pre-Weight line value).

1972

Teuvo Kohonen and **James A. Anderson** each developed a similar network independently of one another. They both used matrix mathematics to describe their ideas but did not realize that what they were doing was creating an array of analog ADALINE circuits.

History of Neural Networks

1982

John Hopfield of Caltech presented a paper to the National Academy of Sciences. His approach was to create more useful machines by using bidirectional lines. Previously, the connections between neurons was only one way.

1982

Joint US-Japan conference on **Cooperative/Competitive Neural Networks**. Japan announced a new Fifth Generation effort on neural networks, and US papers generated worry that the US could be left behind in the field.

1986

Three independent groups of researchers, including **David Rumelhart**, a former member of Stanford's psychology department, came up with similar ideas which are now called back propagation networks.

1997

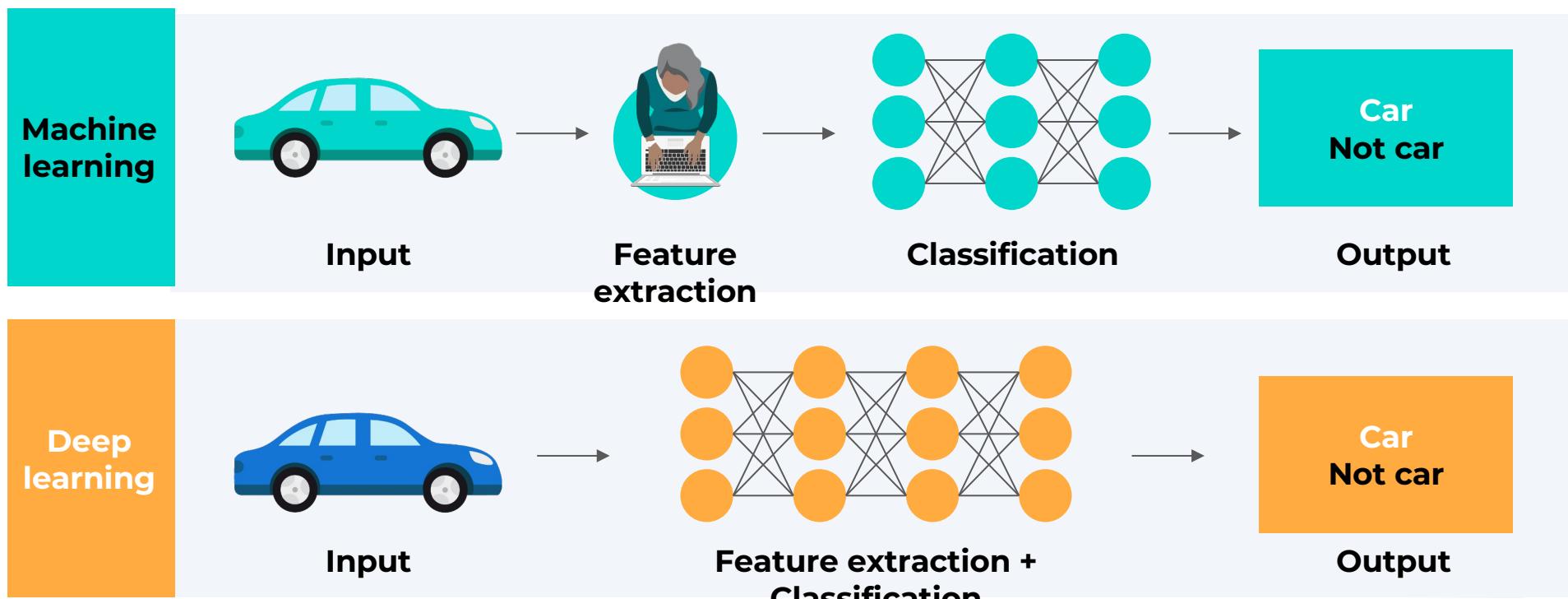
A recurrent neural network framework, LSTM was proposed by **Jürgen Schmidhuber** and **Sepp Hochreiter**.

2000s

Transformers were introduced. Followed by **GANs**, **VAEs**, and **Autoregressive** models which pushed the boundaries of **Generative AI**.

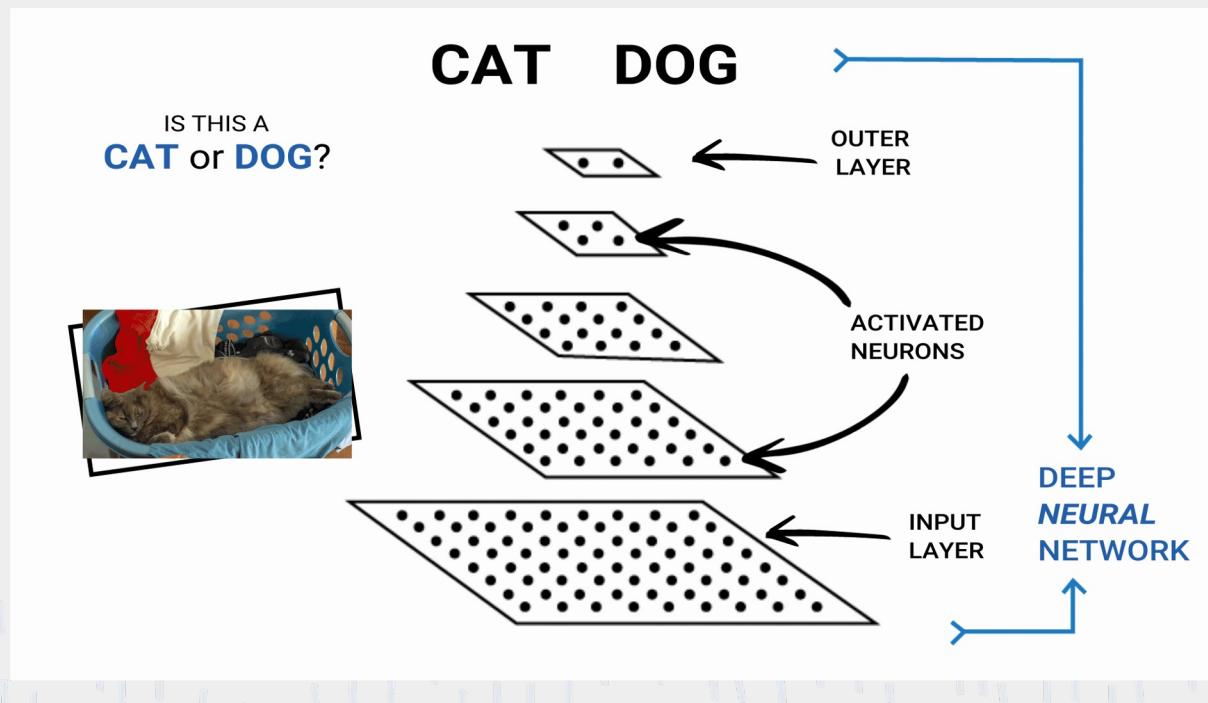
Machine Learning vs. Deep Learning

Deep neural networks are much more effective than traditional machine-learning approaches at discovering nonlinear relationships among data.



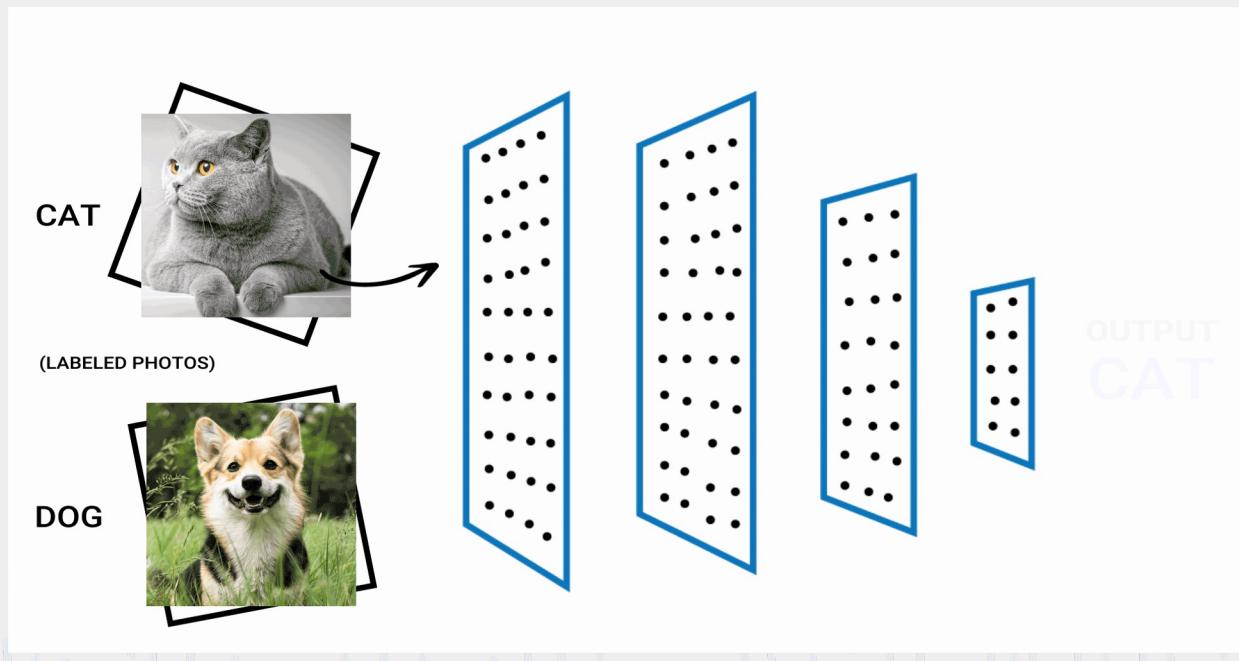
Neural Networks

Neural networks calculate the weights of various input data and pass them to the next layer of neurons. This process continues until the data reaches the output layer, which makes the final decision on the predicted category or numerical value of an instance.



Neural Networks

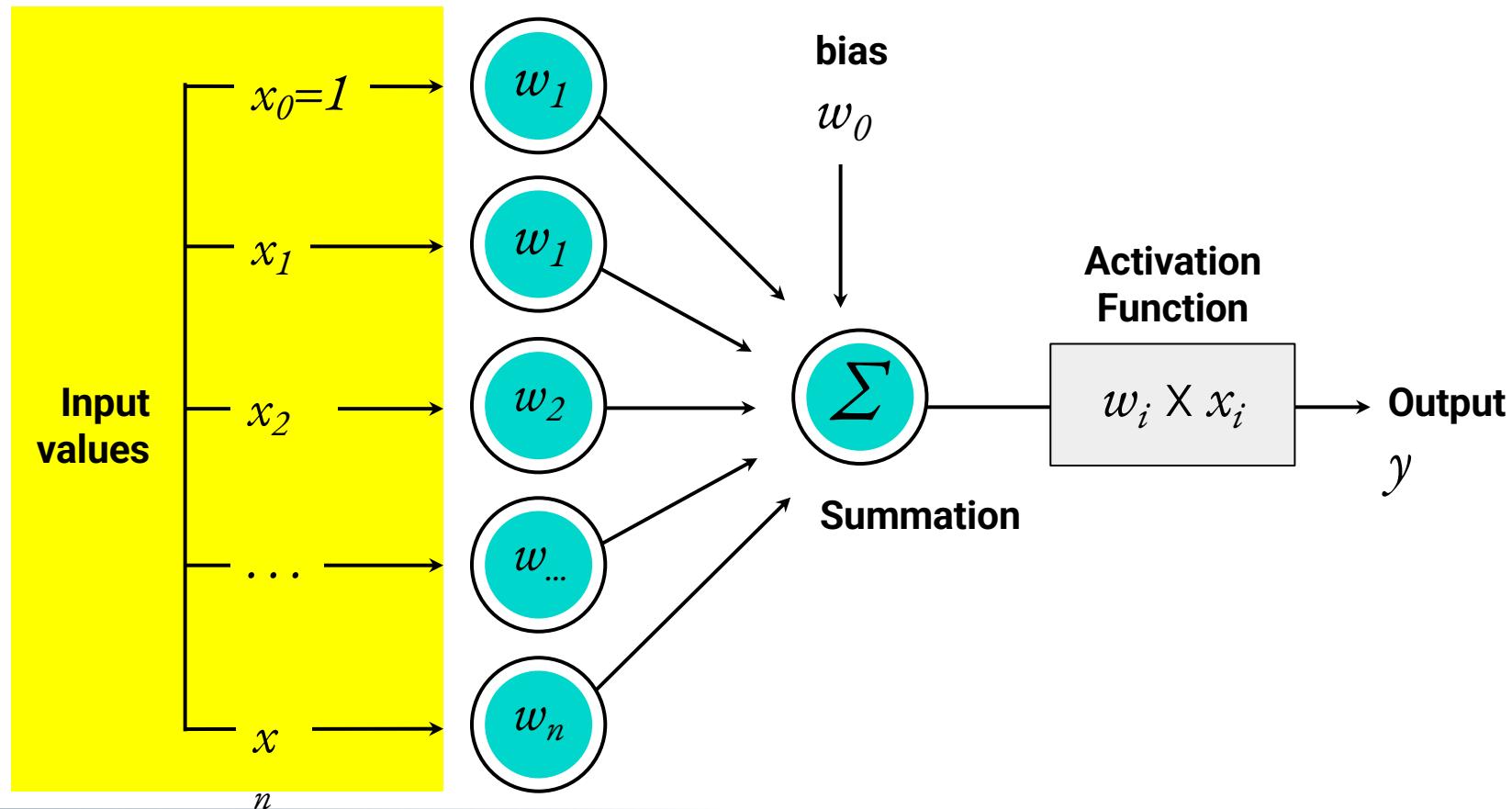
While definitions vary, we can consider neural networks with more than one hidden layer to be deep learning models. The decreasing cost and greater availability of computing power has increased our ability to create and use these models.



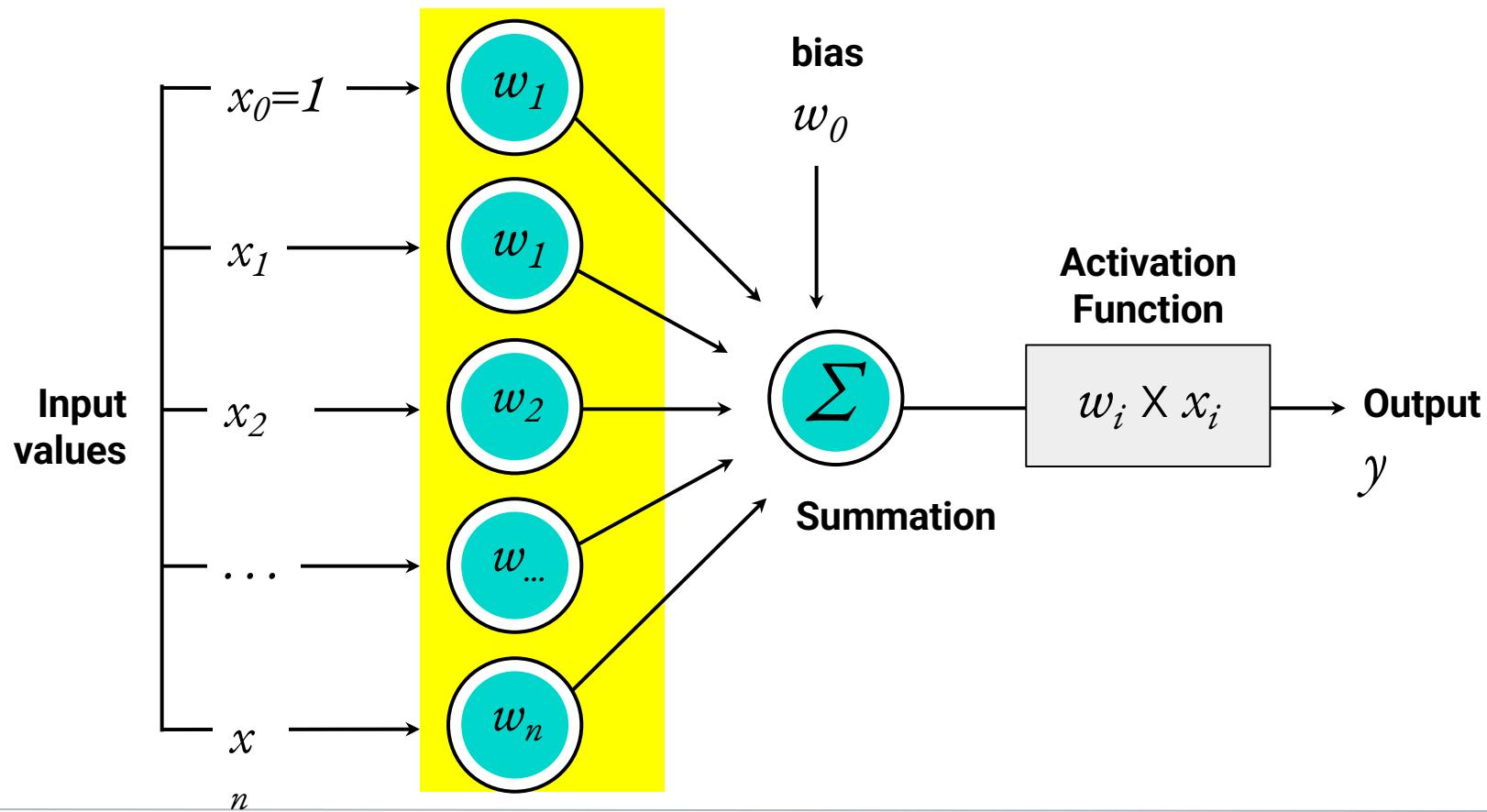


The perceptron model has
four major components.

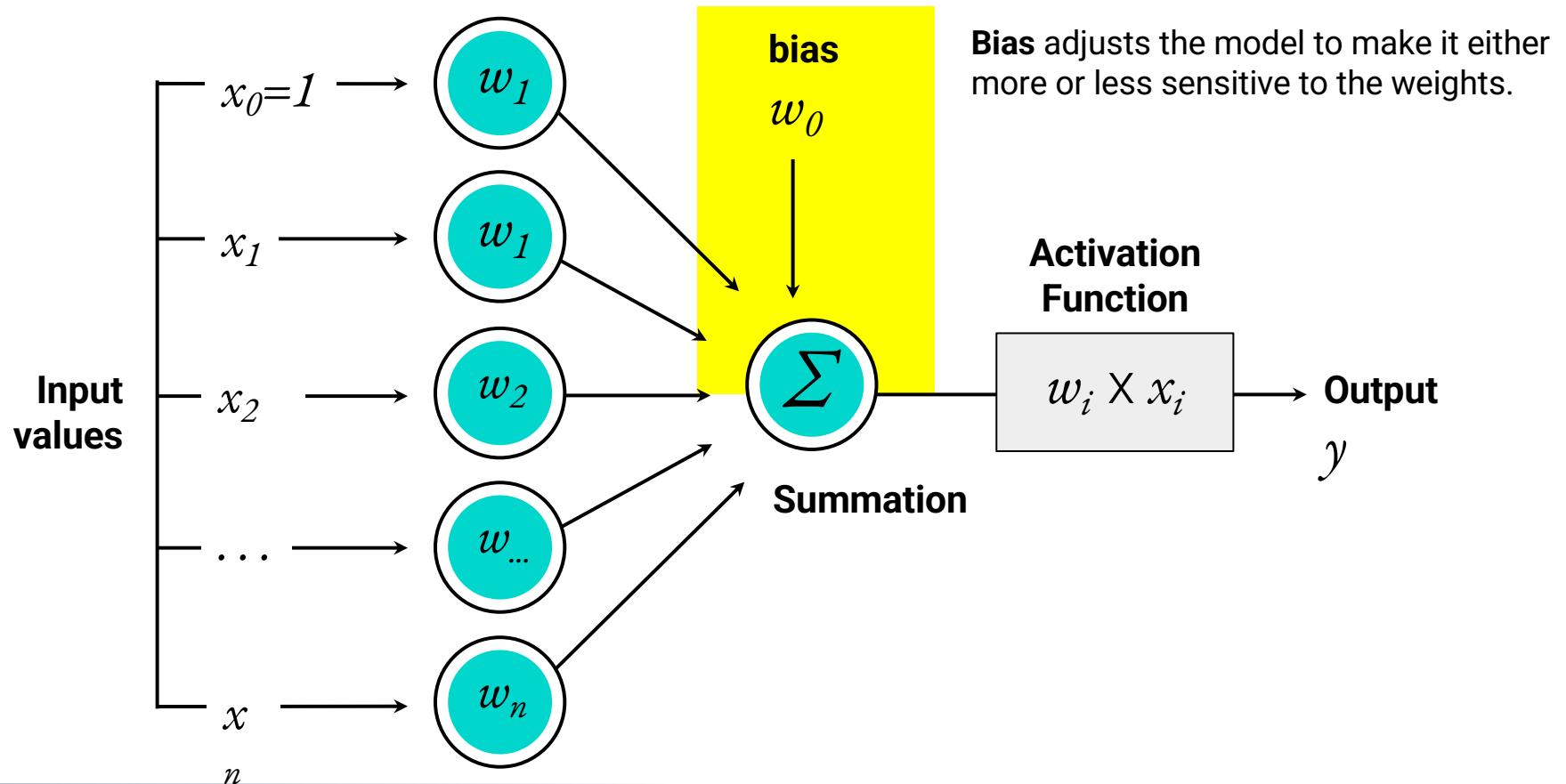
Input Values (shown as x's)



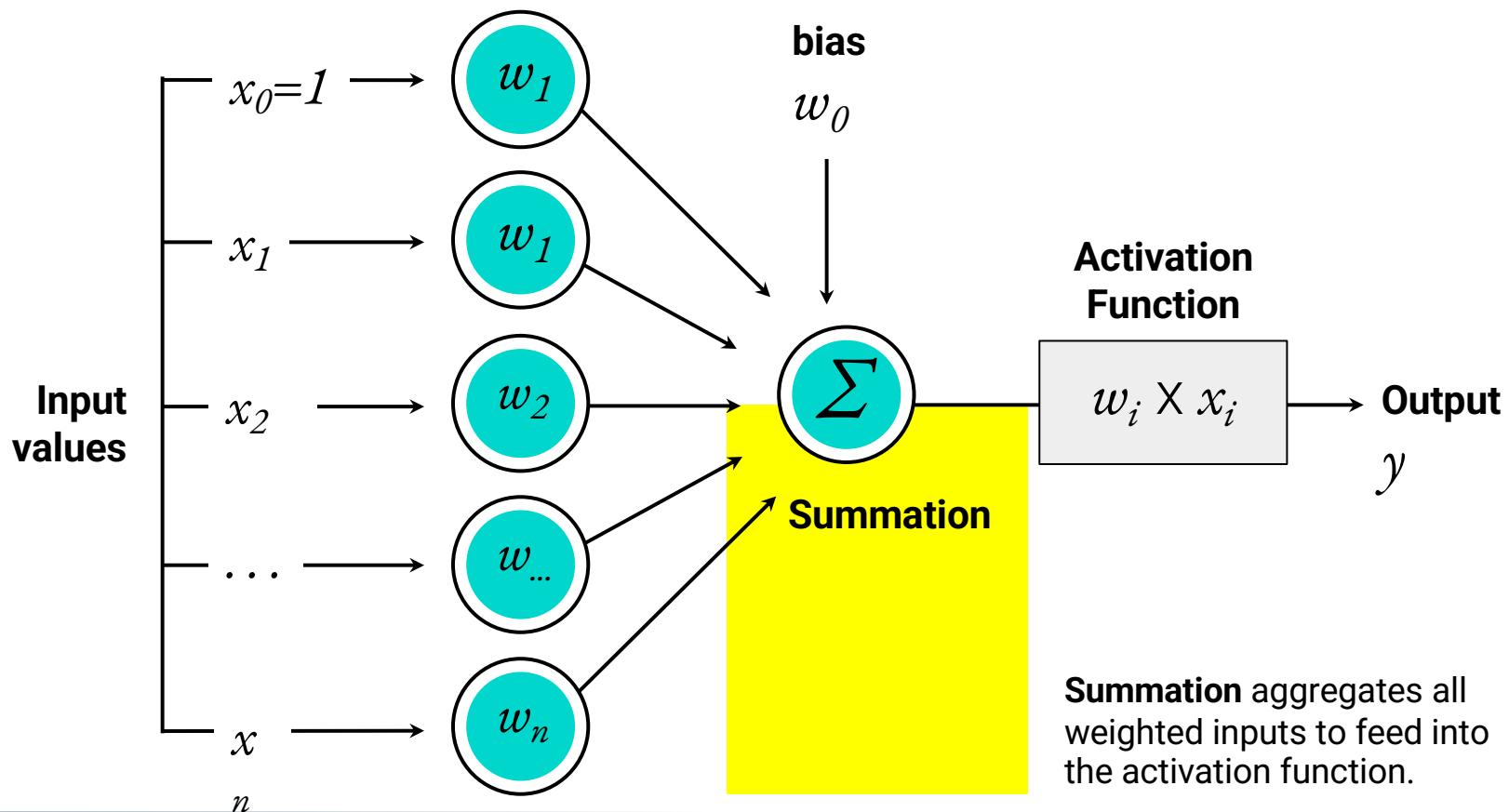
Weight Coefficients (denoted as w's)



A Constant Value Called Bias

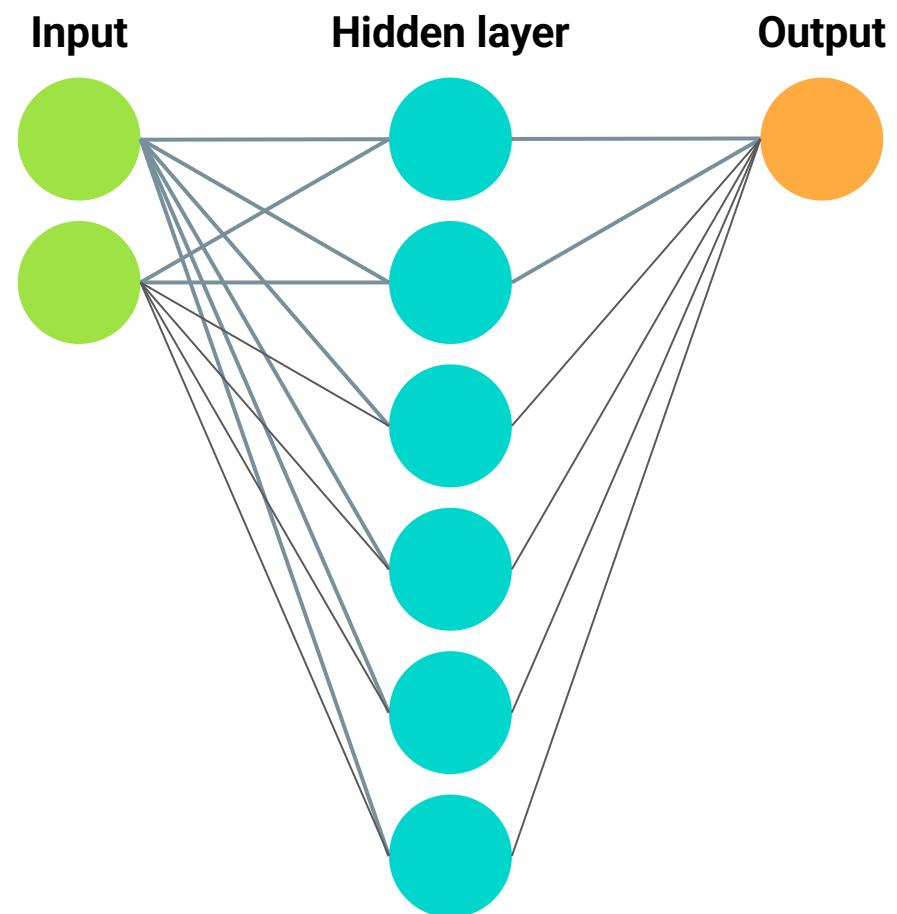


Net Summary Function (the Summation)



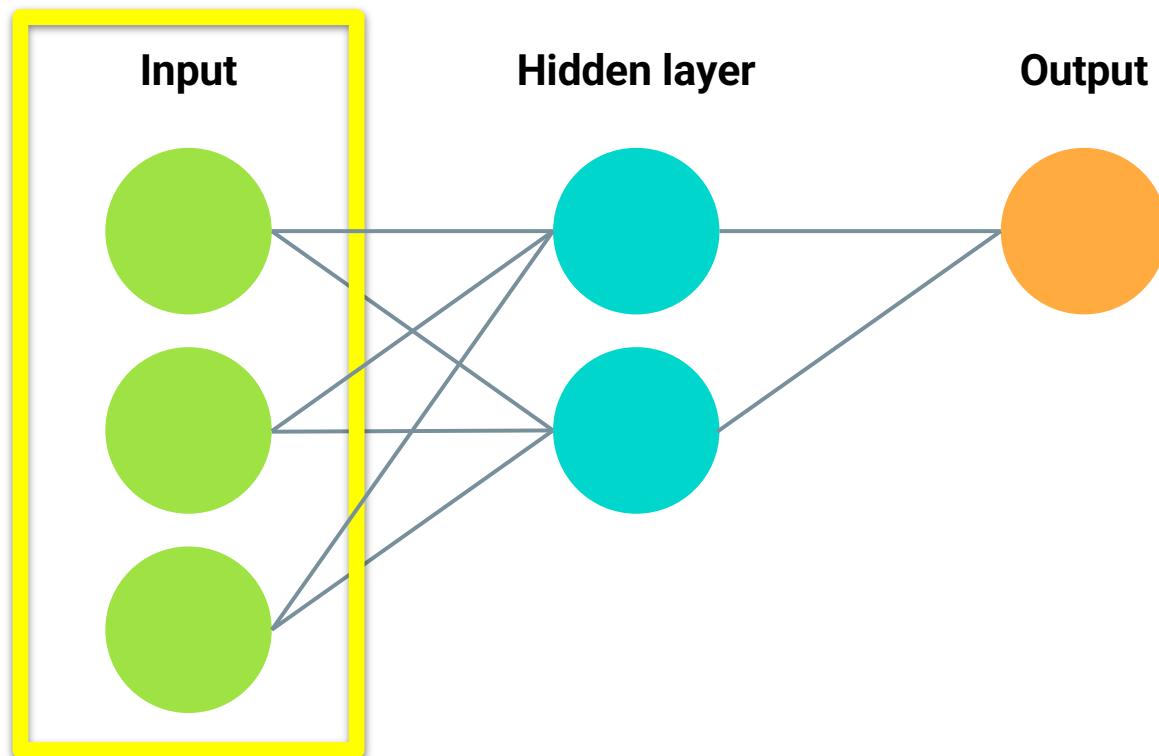
The Neural Network

A modern neural network model is a structure composed of several connected perceptrons that learn from input data to produce an output.



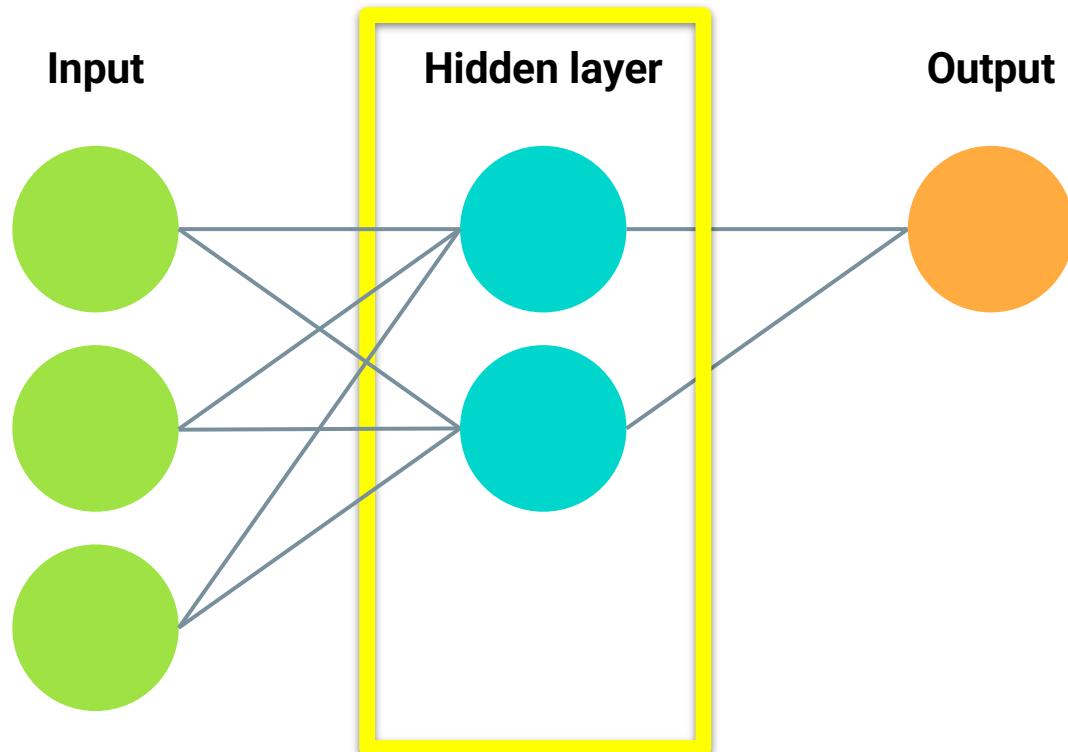
The Structure of a Neural Network

An **input layer** of input values transformed by weight coefficients



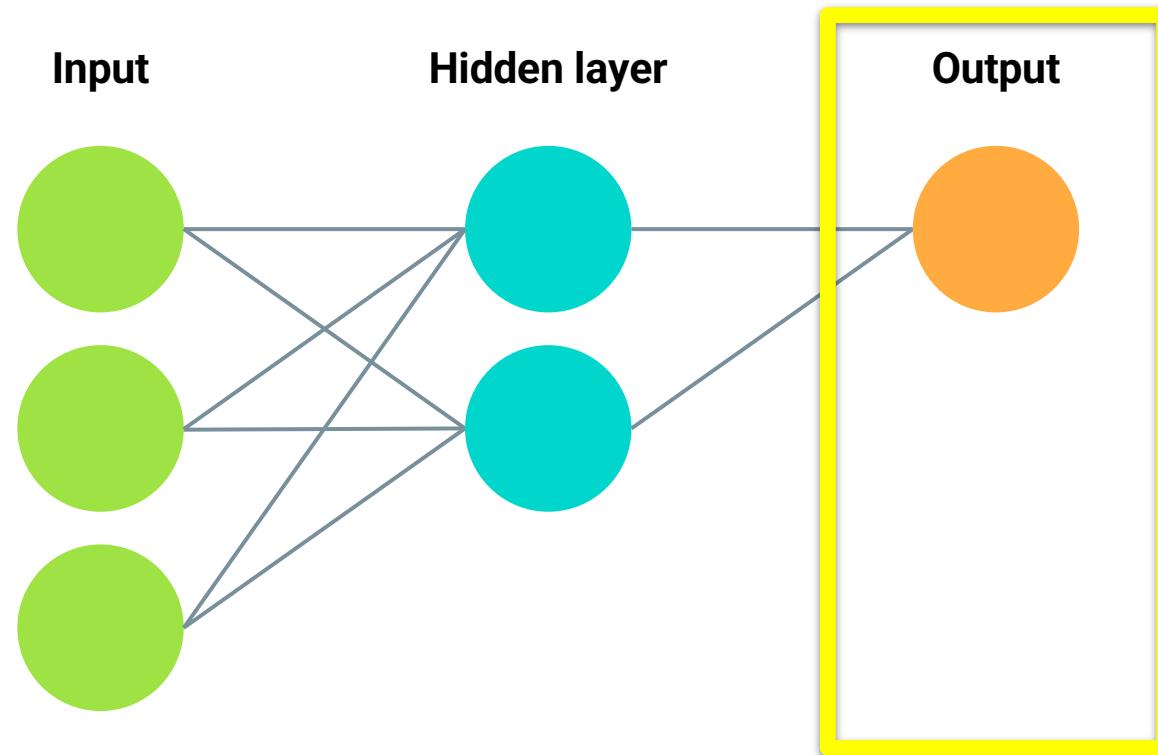
The Structure of a Neural Network

A single **hidden layer** that can contain a single neuron or multiple neurons



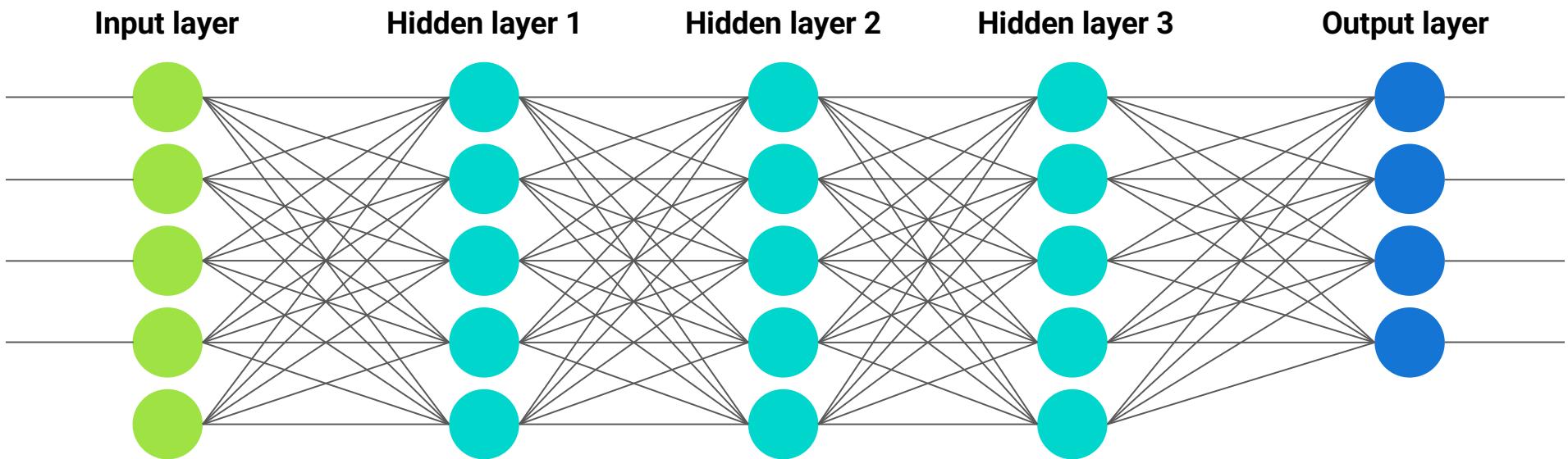
The Structure of a Neural Network

An **output layer** that reports the outcome of the value



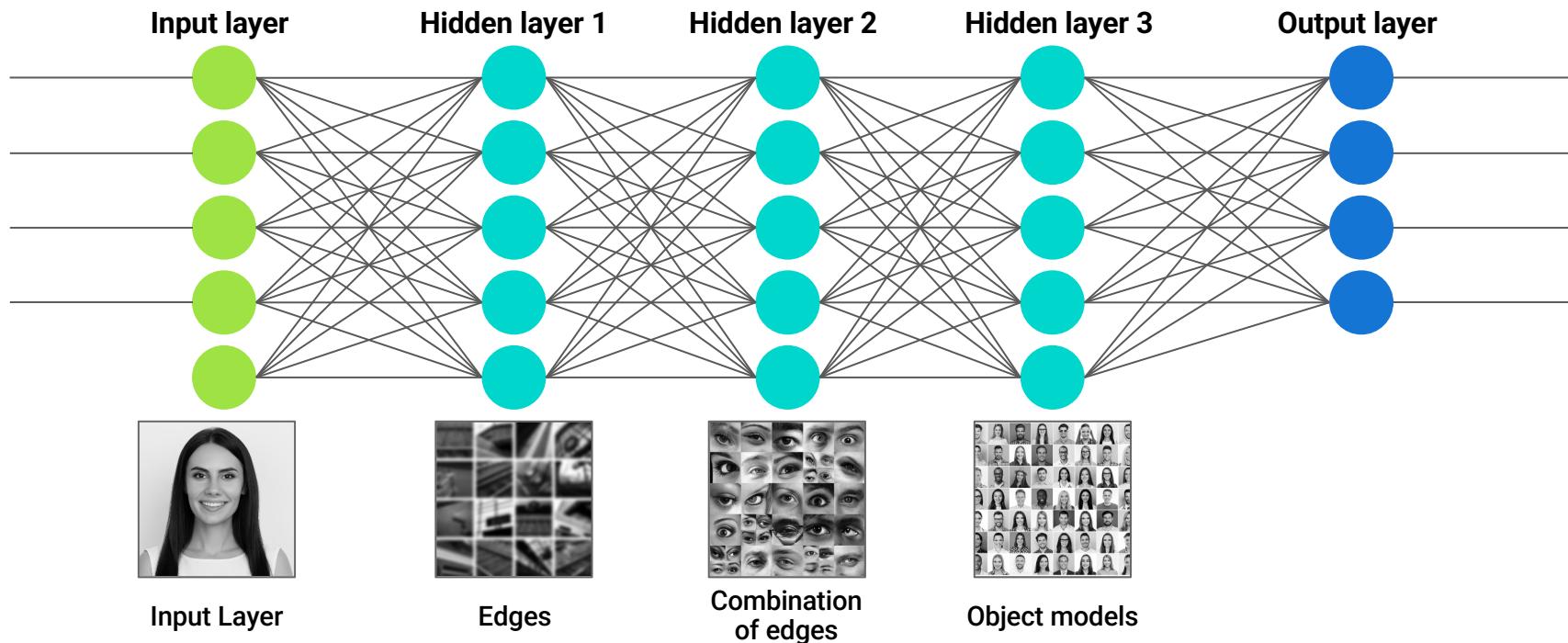
Deep Neural Networks

Generally speaking, deep learning models are neural networks with more than one hidden layer.



Deep Neural Networks

In image recognition, each layer can identify different image features in the process of defining or identifying the image.

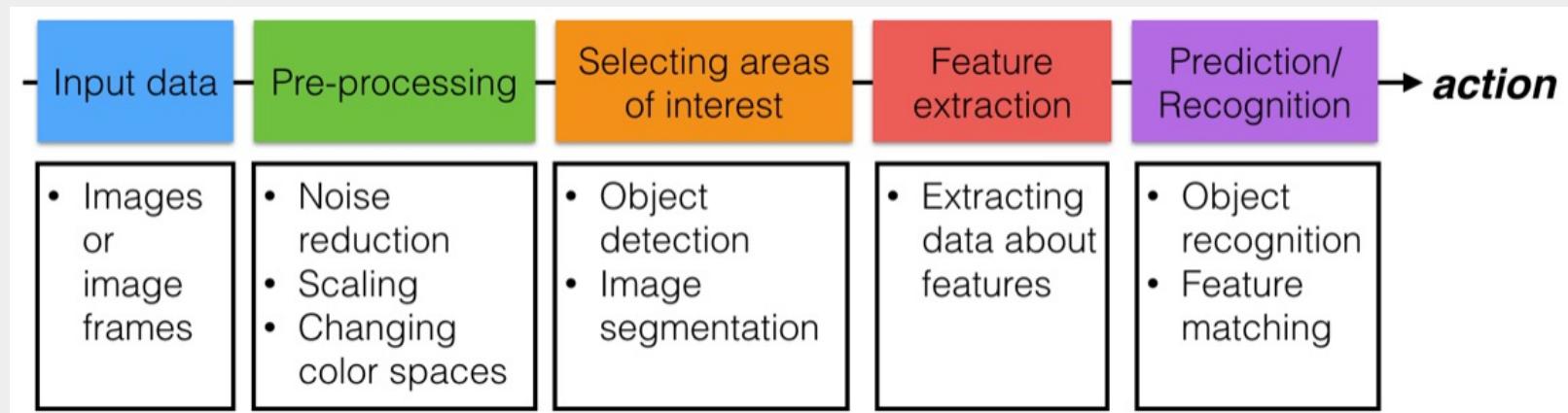


Computer Vision

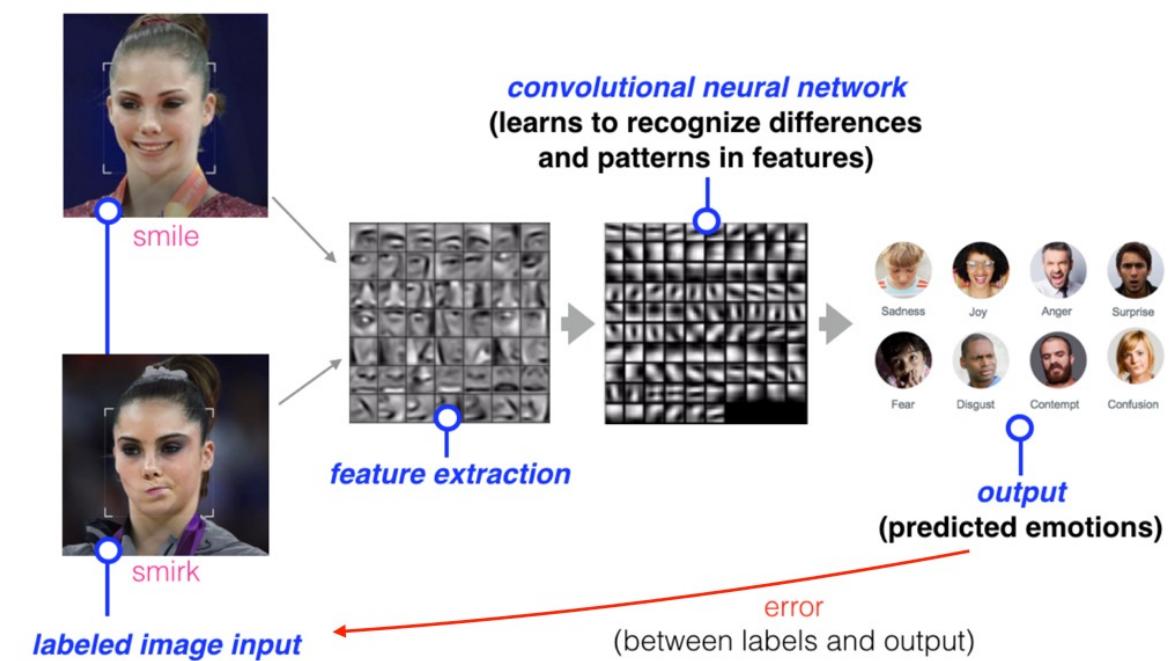
Proprietary and confidential



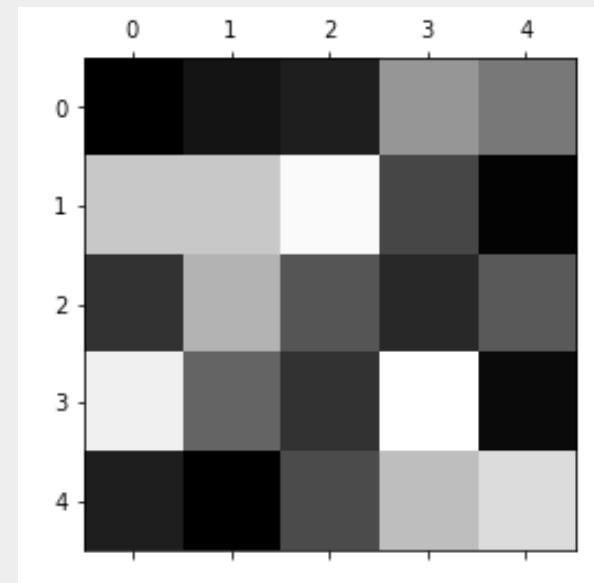
Computer Vision Pipeline



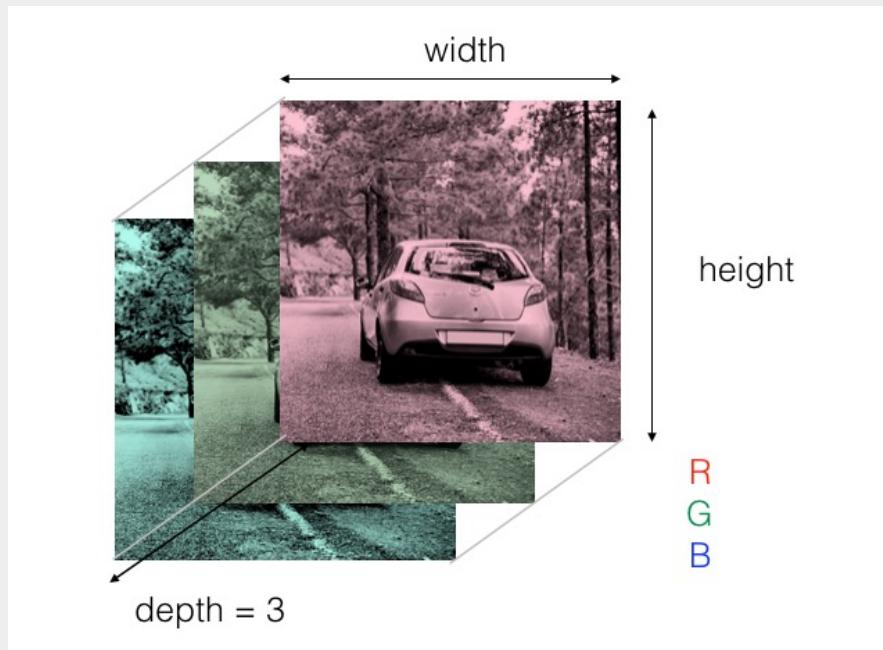
Training a Neural Network



Training a Neural Network



Training a Neural Network



Convolution Kernels – ex Blur



1/9	1/9	1/9
1/9	1/9	1/9
1/9	1/9	1/9

Convolution Kernels – ex Line Detection



-1	-1	-1
2	2	2
-1	-1	-1

Horizontal lines

-1	2	-1
-1	2	-1
-1	2	-1

Vertical lines

-1	-1	2
-1	2	-1
2	-1	-1

45 degree lines

2	-1	-1
-1	2	-1
-1	-1	2

135 degree lines

Convolution Kernels – ex Line Detection



-1	-1	-1
2	2	2
-1	-1	-1

Horizontal lines

-1	2	-1
-1	2	-1
-1	2	-1

Vertical lines

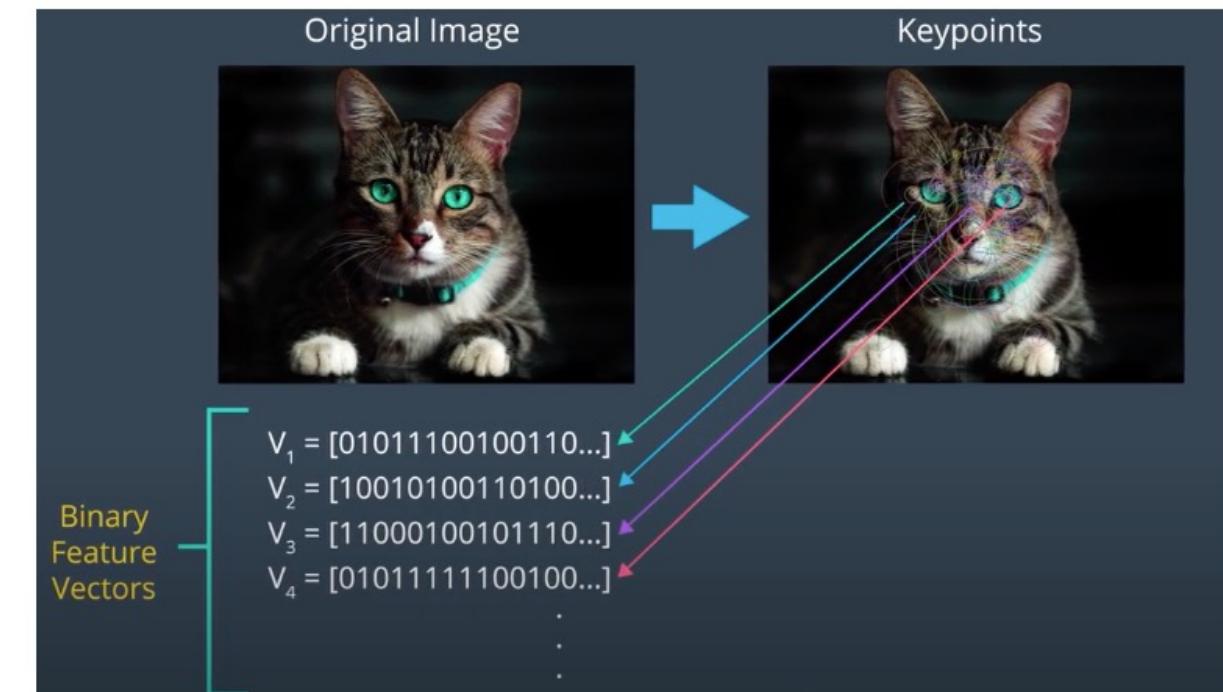
-1	-1	2
-1	2	-1
2	-1	-1

45 degree lines

2	-1	-1
-1	2	-1
-1	-1	2

135 degree lines

Image Classification



Face Detection



Proprietary and confidential

Face Detection – Key Point Detection

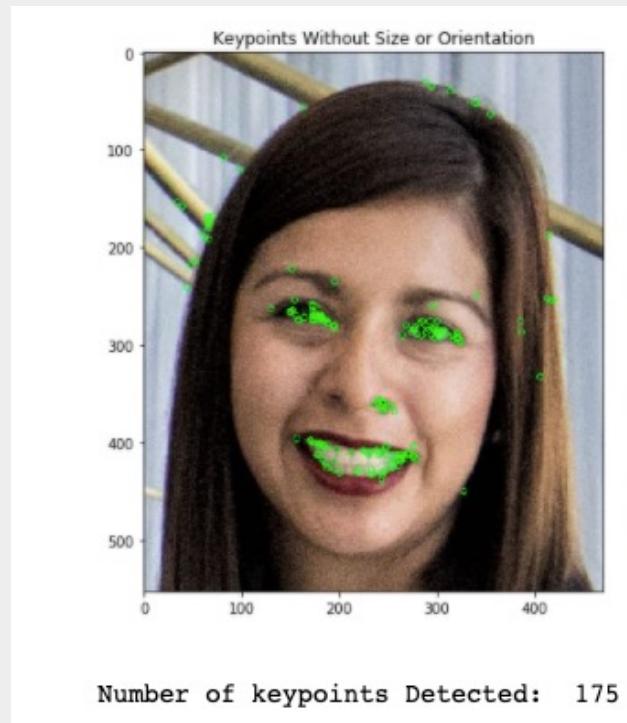
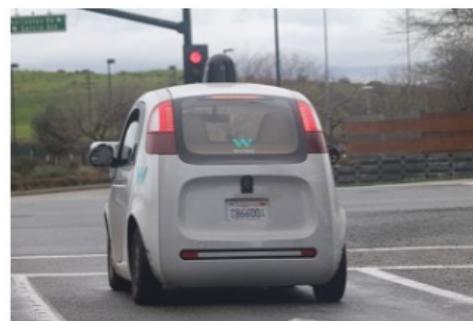


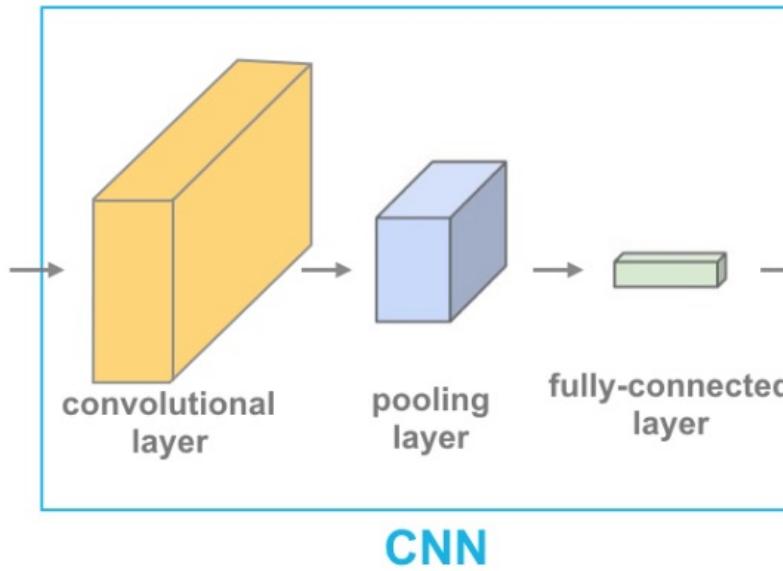
Image Segmentation



Convolutional Neural Networks

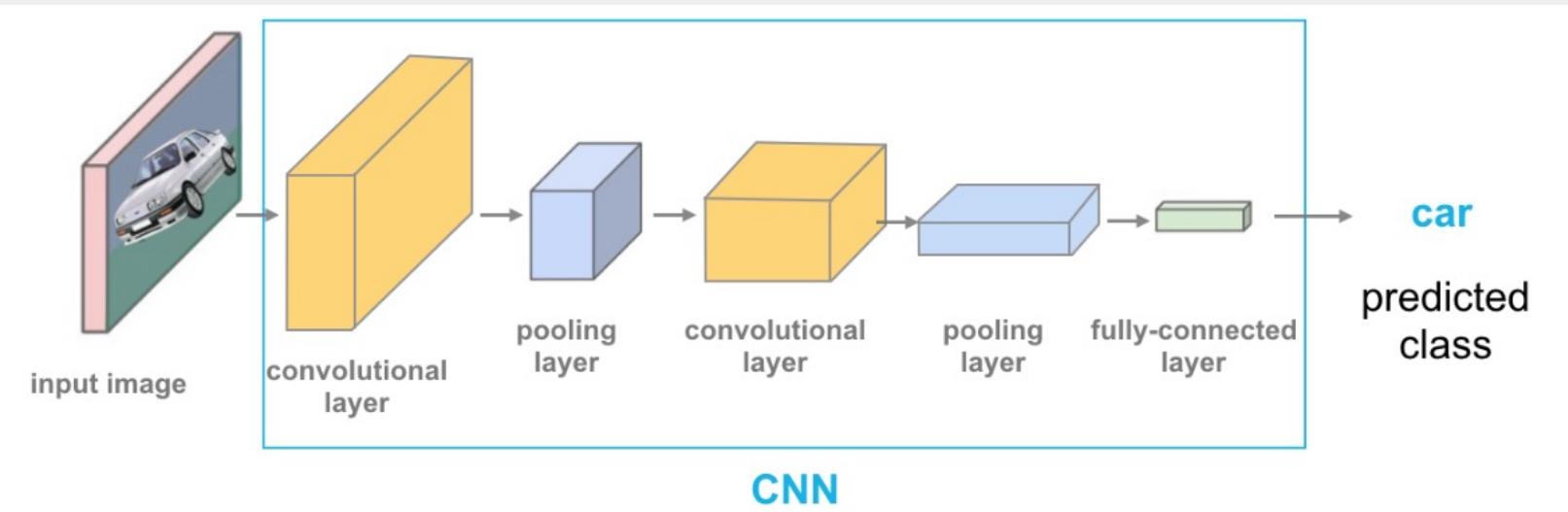


input image



car
predicted
class

Convolutional Neural Networks



Convolutional Neural Networks



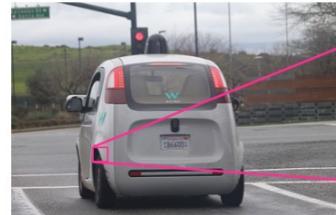
input image

(bus, car, bike, person, ...)

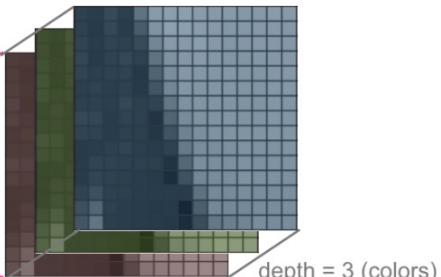
CNN

car

Convolutional Neural Networks



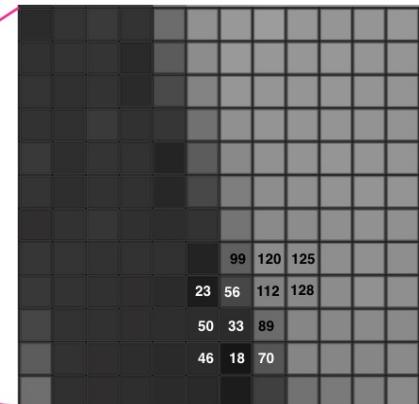
color image



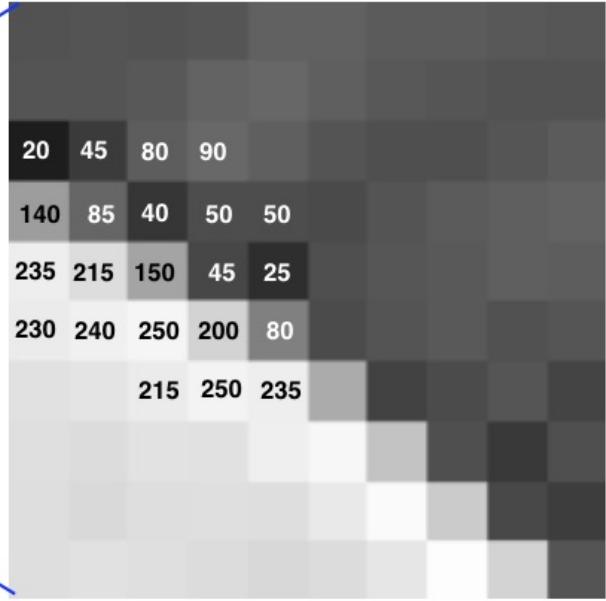
depth = 3 (colors)



grayscale image

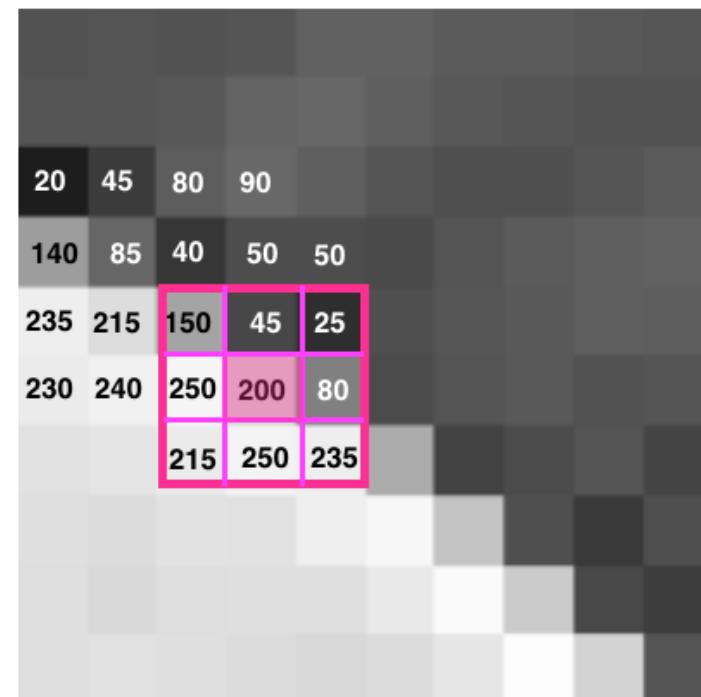


Convolutional Neural Networks



Convolutional Neural Networks

0	-1	0
-1	4	-1
0	-1	0



Convolutional Neural Networks

weights

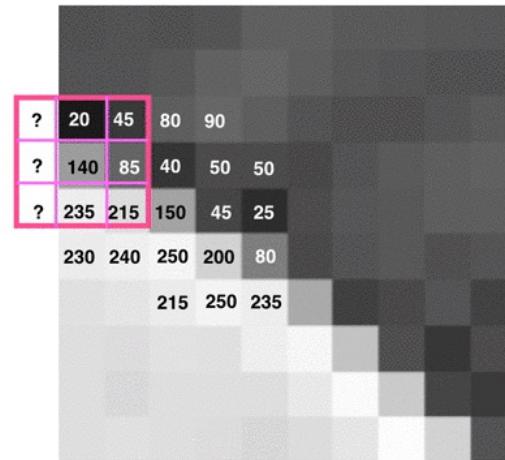
0	-1	0
-1	4	-1
0	-1	0

$$\begin{aligned} & \mathbf{0} * 150 + \mathbf{-1} * 45 + \mathbf{0} * 25 + \\ & \mathbf{-1} * 250 + \mathbf{4} * 200 + \mathbf{-1} * 80 + \\ & \mathbf{0} * 215 + \mathbf{-1} * 250 + \mathbf{0} * 235 \\ & = 175 \end{aligned}$$



Convolutional Neural Networks

0	-1	0
-1	4	-1
0	-1	0



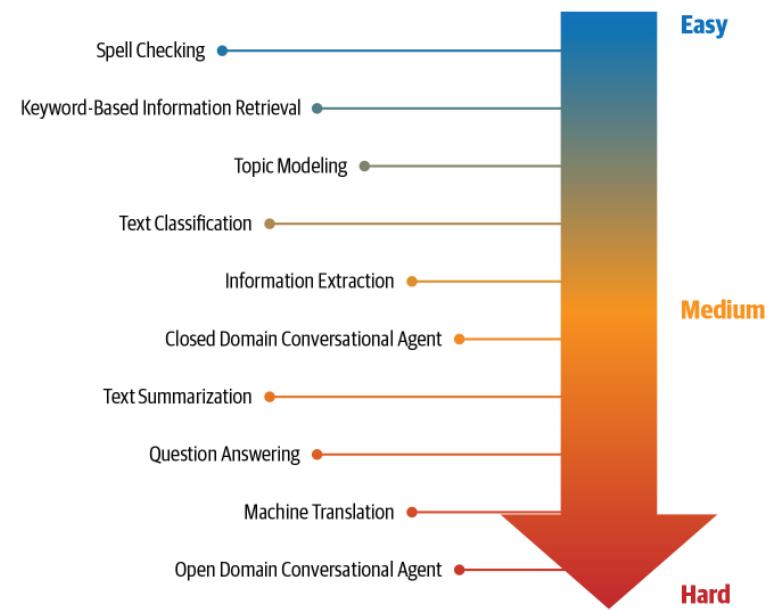
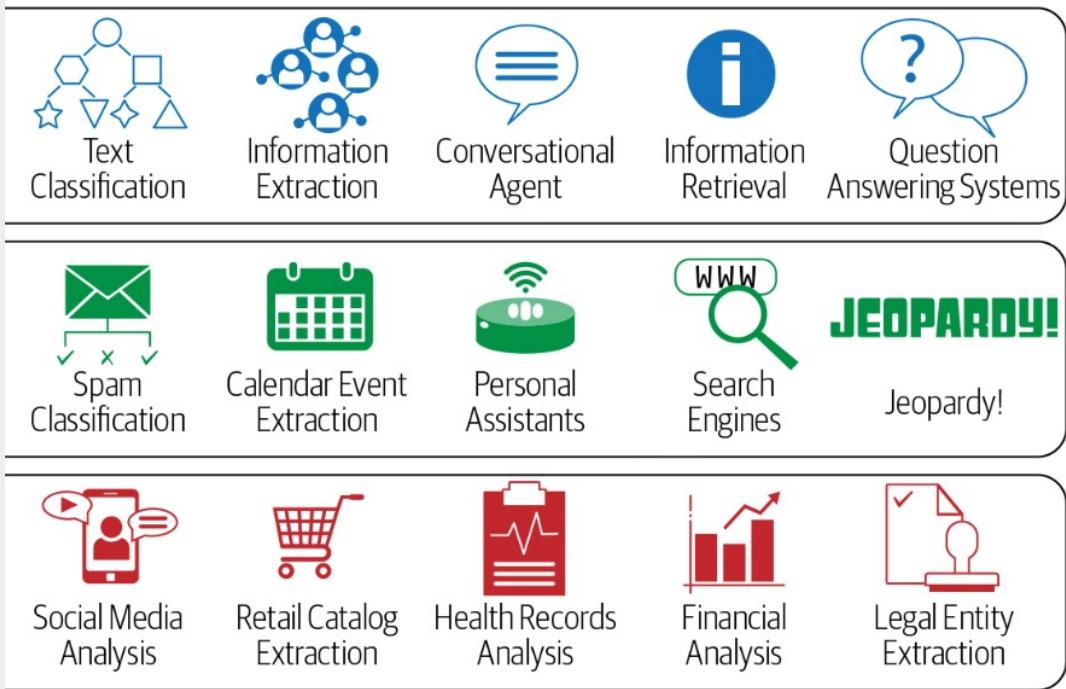
What to do at the edges?

Natural Language Processing

Proprietary and confidential



NLP Tasks



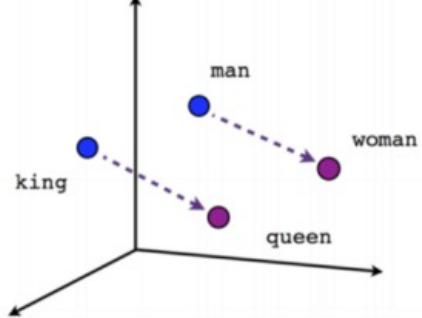
The Bag of Words Representation

I love this movie! It's sweet, but with satirical humor. The dialogue is great and the adventure scenes are fun... It manages to be whimsical and romantic while laughing at the conventions of the fairy tale genre. I would recommend it to just about anyone. I've seen it several times, and I'm always happy to see it again whenever I have a friend who hasn't seen it yet!

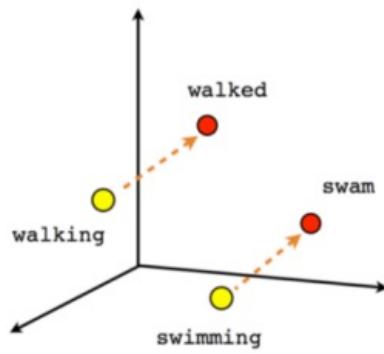
15



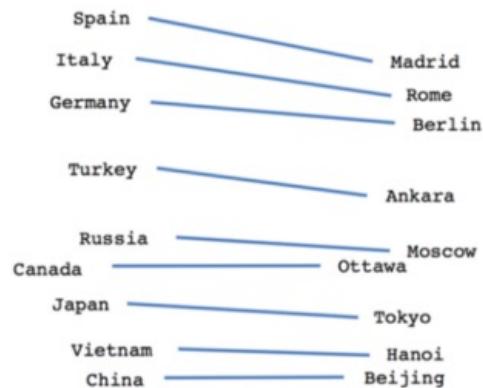
NLP Vector Semantic



Male-Female



Verb tense



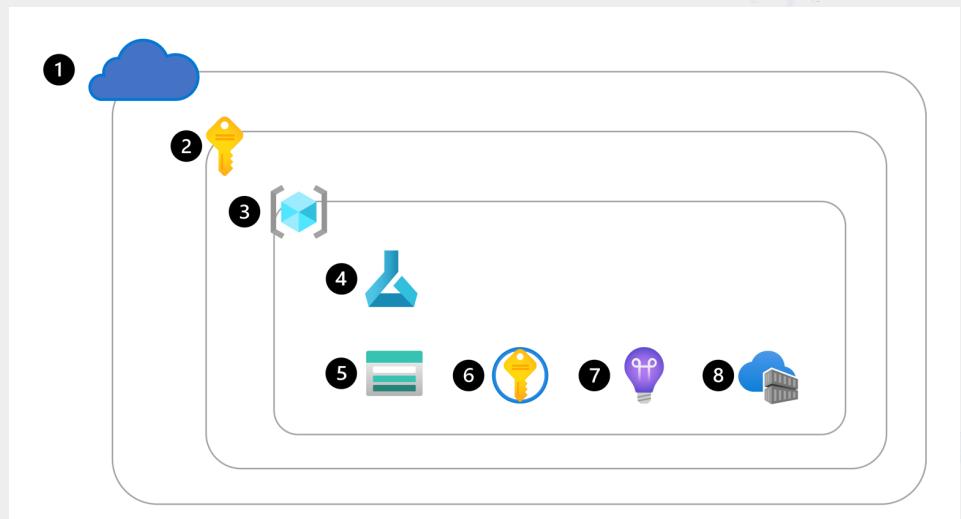
Country-Capital

Azure AI and Machine Learning Services

<https://azure.microsoft.com/en-us/products>

Azure ML Services

- When a **workspace** is provisioned, Azure will automatically create other Azure resources within the same resource group to support the workspace:
- **Azure Storage Account**: To store files and notebooks used in the workspace, and to store metadata of jobs and models.
- **Azure Key Vault**: To securely manage secrets such as authentication keys and credentials used by the workspace.
- **Application Insights**: To monitor predictive services in the workspace.
- **Azure Container Registry**: Created when needed to store images for Azure Machine Learning environments.



Azure ML Compute Instances

A compute instance is a fully managed cloud-based workstation optimized for your machine learning development environment.

- **Compute instances:** Similar to a virtual machine in the cloud, managed by the workspace. Ideal to use as a development environment to run (Jupyter) notebooks.
- **Compute clusters:** On-demand clusters of CPU or GPU compute nodes in the cloud, managed by the workspace. Ideal to use for production workloads as they automatically scale to your needs.
- **Kubernetes clusters:** Allows you to create or attach an Azure Kubernetes Service (AKS) cluster. Ideal to deploy trained machine learning models in production scenarios.
- **Attached computes:** Allows you to attach other Azure compute resources to the workspace, like Azure Databricks or Synapse Spark pools.
- **Serverless compute:** A fully managed, on-demand compute you can use for training jobs.

Datastores

The workspace doesn't store any data itself. Instead, all data is stored in **datastores**, which are references to Azure data services.

- **workspaceartifactstore**: Connects to the `azureml` container of the Azure Storage account created with the workspace. Used to store compute and experiment logs when running jobs.
- **workspaceworkingdirectory**: Connects to the file share of the Azure Storage account created with the workspace used by the **Notebooks** section of the studio. Whenever you upload files or folders to access from a compute instance, it's uploaded to this file share.
- **workspaceblobstore**: Connects to the Blob Storage of the Azure Storage account created with the workspace. Specifically the `azureml-blobstore-...` container. Set as the default datastore, which means that whenever you create a data asset and upload data, it's stored in this container.
- **workspacefilestore**: Connects to the file share of the Azure Storage account created with the workspace. Specifically the `azureml-filestore-...` file share.

ML Assets

- Models
- Environments
- Data
- Components

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Authoring within ML Workspace

Authoring

Notebooks

Automated ML

Designer

Prompt flow

Tracing PREVIEW

Designer

Azure AI | Machine Learning Studio

Default Directory > dp100-atwan > Designer > Authoring

Undo Redo Validate Show lineage Clone AutoSave

Configure & Submit

Save Pipeline interface

Regression - Automobile Price Prediction (Basic)

Tags: All Add filter

Data Component

95 +

All workspaces Home Model catalog

Authoring Notebooks Automated ML

Designer Prompt flow Tracing PREVIEW

Assets Data Jobs

Components Data Input and Output (3) Recommendation (5)

Pipelines

Environments

Models

Endpoints

Manage Compute Monitoring

Data Labeling

Linked Services PREVIEW

Connections PREVIEW

Automobile price data (Raw)

Select Columns in Dataset

Clean Missing Data

Split Data

Linear Regression linear_regression

Train Model train model

Score Model score model

Evaluate Model evaluate model

Proprietary and confidential

PLURALSIGHT

The screenshot shows the Azure Machine Learning Studio Designer interface. On the left, there's a navigation sidebar with various sections like 'All workspaces', 'Authoring', 'Designer' (which is selected), and 'Components'. The main area displays a pipeline titled 'Regression - Automobile Price Prediction (Basic)'. The pipeline starts with a 'Raw' dataset, which undergoes 'Select Columns in Dataset' (excluding normalized losses with many missing values), 'Clean Missing Data' (removing missing value rows), and 'Split Data' (splitting the dataset into training set (0.7) and test set (0.3)). The training set feeds into a 'Linear Regression' component ('linear_regression'). The output of this component is an 'Untrained model', which then goes into a 'Train Model' component ('train model'). The output of 'Train Model' is a 'Trained model', which then feeds into a 'Score Model' component ('score model'). Finally, the output of 'Score Model' goes into an 'Evaluate Model' component ('evaluate model'). There are also intermediate datasets labeled 'Resampled dataset', 'Cleaned dataset', 'Training tran...', 'Resulted dataset...', 'Results dataset...', 'Trained model', 'Dataset', 'Scored dataset...', and 'Scored dataset...'. The interface includes standard tools at the top: Undo, Redo, Validate, Show lineage, Clone, AutoSave, Configure & Submit, Save, and Pipeline interface. The bottom of the interface has a Navigator, zoom controls (100%), and search/filter icons.

Designer

Azure AI | Machine Learning Studio

Default Directory > dp100-atwan > Designer > Authoring

Regression - Automobile Price Prediction (Basic)

Configure & Submit

Save Pipeline interface

Tags: All Add filter

Data Component

95 +

Sample data (16)

Data Transformation (19)

Computer Vision (6)

Model Scoring & Evaluation (6)

Machine Learning Algorithms (19)

Text Analytics (7)

Python Language (2)

Data Input and Output (3)

Recommendation (5)

R Language (1)

Feature Selection (2)

Anomaly Detection (2)

Statistical Functions (1)

Model Training (4)

Web Service (2)

All workspaces Home Model catalog Authoring Notebooks Automated ML Designer Prompt flow Tracing PREVIEW Assets Data Jobs Components Pipelines Environments Models Endpoints Manage Compute Monitoring Data Labeling Linked Services PREVIEW Connections PREVIEW

Automobile price data (Raw)

Select Columns in Dataset

Clean Missing Data

Linear Regression linear_regression

Train Model train model

Score Model score model

Evaluate Model evaluate model

Proprietary and confidential

PLURALSIGHT

The screenshot shows the Azure Machine Learning Studio Designer interface. On the left, there's a navigation sidebar with various sections like Home, Model catalog, Authoring, Components, Pipelines, and so on. The 'Designer' section is currently selected. The main area displays a pipeline titled 'Regression - Automobile Price Prediction (Basic)'. The pipeline starts with 'Automobile price data (Raw)', which is processed by 'Select Columns in Dataset' (excluding normalized losses with many missing values) and 'Clean Missing Data' (removing missing value rows). The resulting 'Cleaned data...' is then split into training and test sets by 'Split Data'. The 'Untrained-model' output from 'Linear Regression' is used as input for 'Train Model'. The 'Trained-model' output from 'Train Model' is then used for 'Score Model' and 'Evaluate Model'. The 'Scored dataset...' and 'Evaluated dataset...' outputs are shown at the bottom. The pipeline interface includes standard tools like Undo, Redo, Validate, Show lineage, Clone, AutoSave, and a Configure & Submit button.

Designer

Undo Redo Validate Show lineage ...

Save Pipeline interface

...

Linear Regression linear_regression

Untrained model

Train Model train model

Trained model

Score Model score m

Score

Evaluate evaluate

Navigator 100% Back Next Close

Proprietary and confidential

PLURALSIGHT

Set up pipeline job

Basics

Experiment name Select existing Create new

Existing experiment *

Job display name

Job description

Job tags Name : Value

Review + Submit Back Next Close

Designer – A Compute Cluster

The screenshot shows the Azure Machine Learning Designer interface. On the left, a pipeline diagram is visible, consisting of several components connected by arrows:

- A "Linear Regression" component with the ID "linear_regression".
- An "Untrained model" output from the Linear Regression component.
- A "Train Model" component with the ID "train model".
- An "Untrained model" input to the Train Model component, which receives the "Untrained model" from the Linear Regression component.
- A "Trained model" output from the Train Model component.
- A "Score Model" component with the ID "score m".
- An "Untrained model" input to the Score Model component, which receives the "Trained model" from the Train Model component.
- A "Score" output from the Score Model component.
- An "Evaluate" component with the ID "evaluate".
- An "Untrained model" input to the Evaluate component, which receives the "Score" output from the Score Model component.
- A "Score" output from the Evaluate component.

The interface has a top navigation bar with "Undo", "Redo", "Validate", "Show lineage", and other options. Below the navigation is a toolbar with "Save" and "Pipeline interface" buttons. The main area is titled "Set up pipeline job" and contains the following steps:

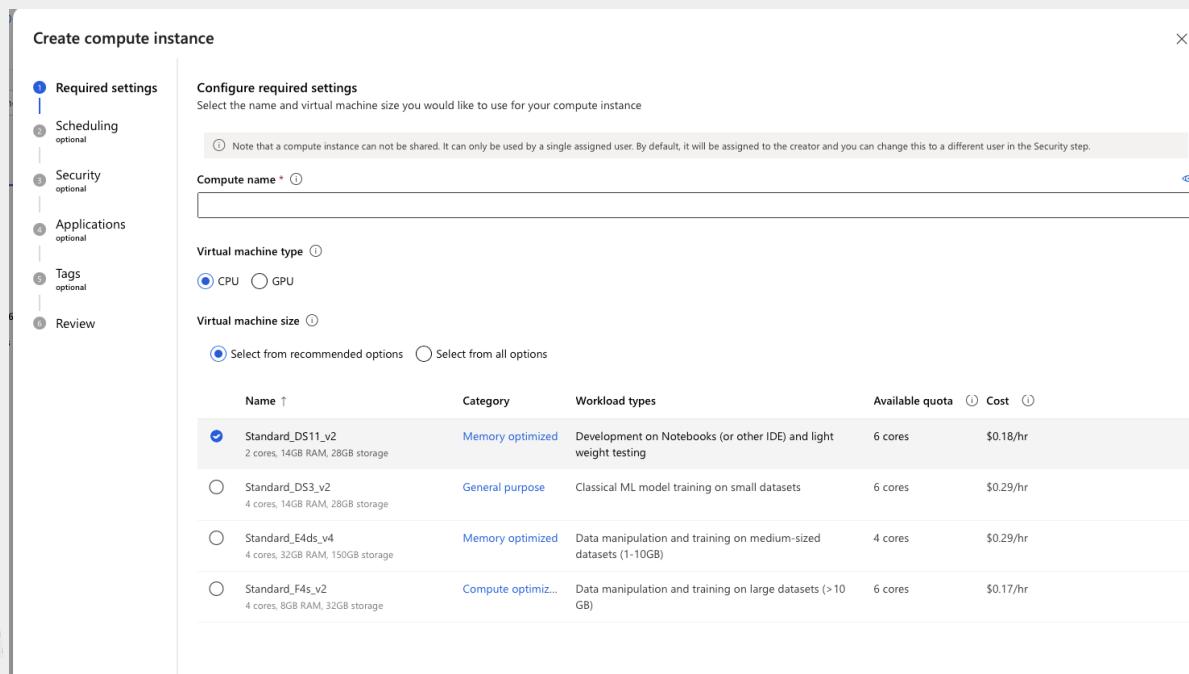
- Basics**: Completed.
- Inputs & outputs**: Completed.
- Runtime settings**: In progress. A red error message says "Please select a default compute to run a pipeline." A dropdown menu under "Select compute type" shows "Compute cluster".
- Review + Submit**: Not yet started.

On the right side of the dialog, there is a preview window showing the "Runtime settings" step with the same error message and dropdown menu. At the bottom of the dialog are buttons for "Review + Submit", "Back", "Next", and "Close".

At the bottom left of the interface, there is a watermark that reads "Proprietary and confidential". At the bottom right, the Pluralsight logo is present.

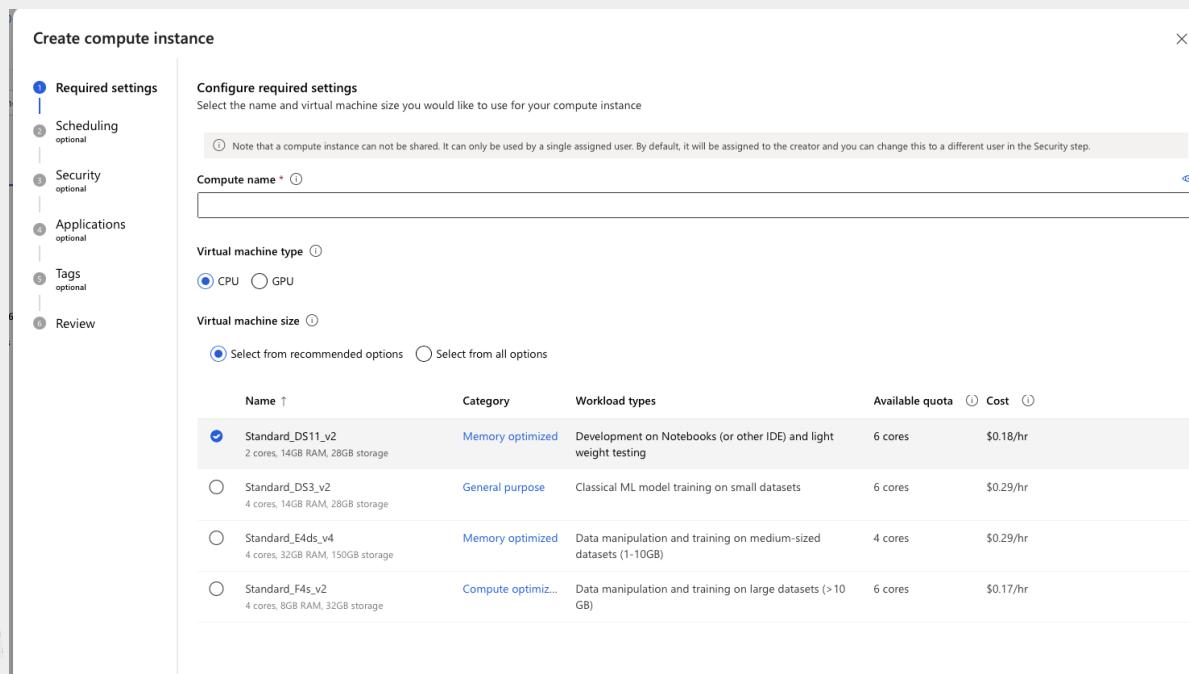
What is a Computer Instance in ML Workspace

- A compute instance is a fully managed cloud-based workstation optimized for your machine learning development environment.



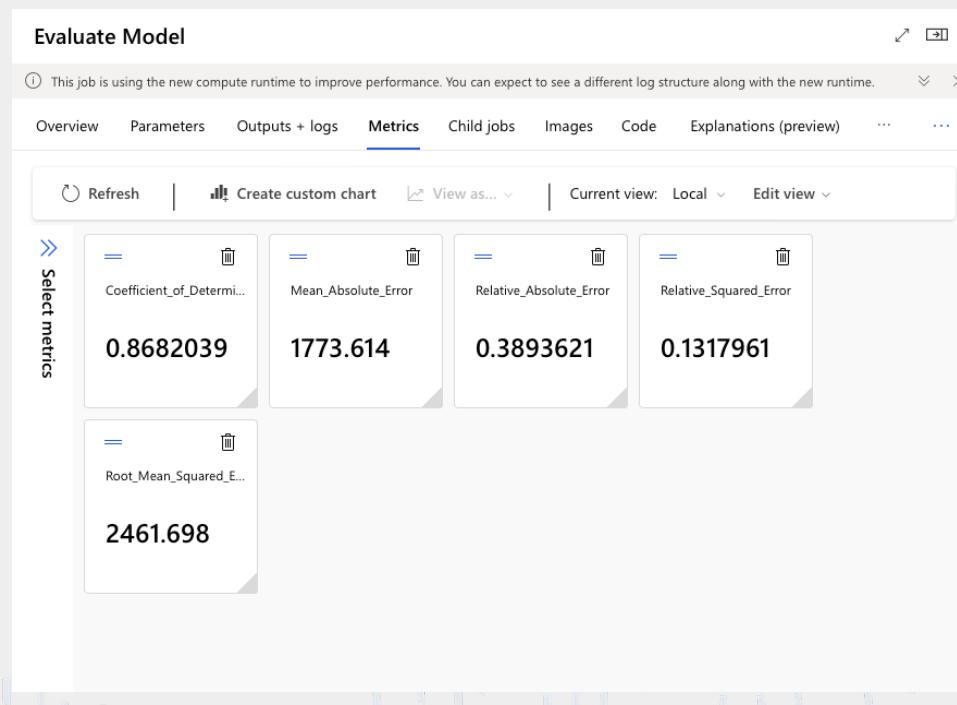
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- A compute instance is a fully managed cloud-based workstation optimized for your machine learning development environment.



What is a Computer Instance in ML Workspace

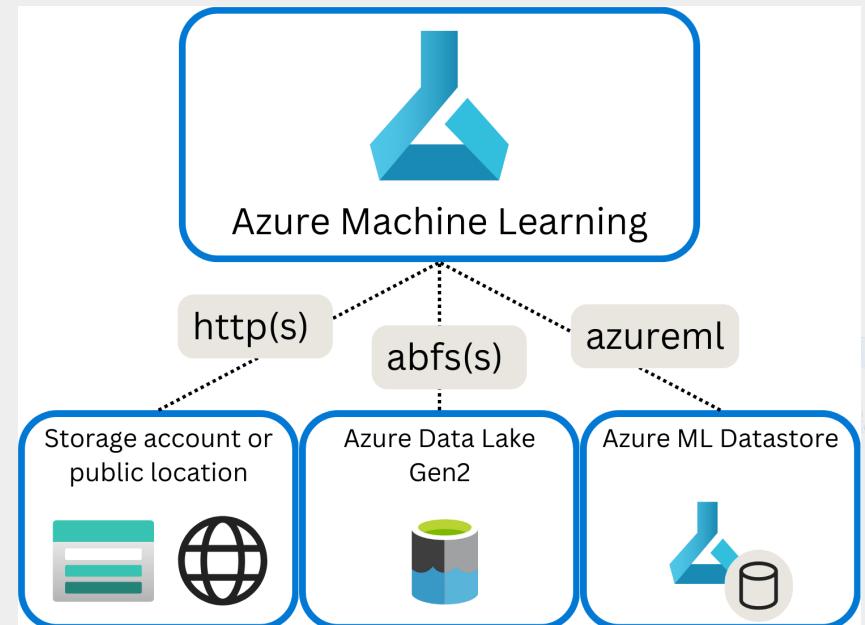
- A compute instance is a fully managed cloud-based workstation optimized for your machine learning development environment.



Data in Azure ML Workspace

To find and access data in Azure Machine Learning, you can use **Uniform Resource Identifiers (URIs)**.

- **http(s)**: Use for data stores publicly or privately in an Azure Blob Storage or publicly available http(s) location.
- **abfs(s)**: Use for data stores in an Azure Data Lake Storage Gen 2.
- **azureml**: Use for data stored in a datastore.



Thank you!

If you have any additional questions, please ask! If



PLURALSIGHT