# Technology Lab: Using Al Frameworks in Jupyter Notebook

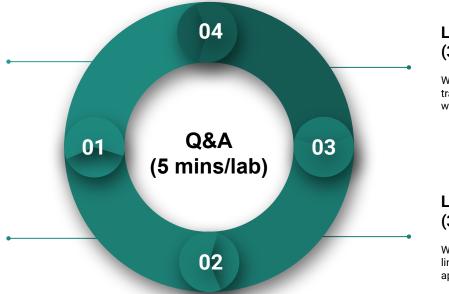
#### Zhenhua He 02/21/2023





High Performance Research Computing DIVISION OF RESEARCH

# **AI Tech Labs**



### Lab IV. Deep Learning (30 minutes)

We will learn how to use Keras to build and train a simple image classification model with deep neural network (DNN).

### Lab III Machine Learning (30 minutes)

We will learn to use scikit-learn library for linear regression and classification applications.

Figure 1. Structure of the AI Technology Labs.

#### Lab I. JupyterLab (30 mins)

We will load required modules with Jupyter Lmode extension and run JupyterLab on HPRC portal.

#### Lab II. Data Exploration (30 mins)

We will go through some examples with two popular Python libraries: Pandas and Matplotlib for data exploration.

# Lab I. JupyterLab



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		Output View ×	lorenz.py ×									
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# L1 - Resources

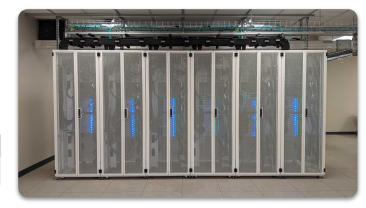
- Texas A&M High Performance Research Computing (HPRC)
- FASTER Quick Start Guide
- ACES Phase I Guide
- ACCESS Documentation
- FASTER Portal (TAMU)
- FASTER Portal (ACCESS)
- HPRC YouTube Channel
- help@hprc.tamu.edu

# Getting Started with FASTER and ACES

# **FASTER Cluster**

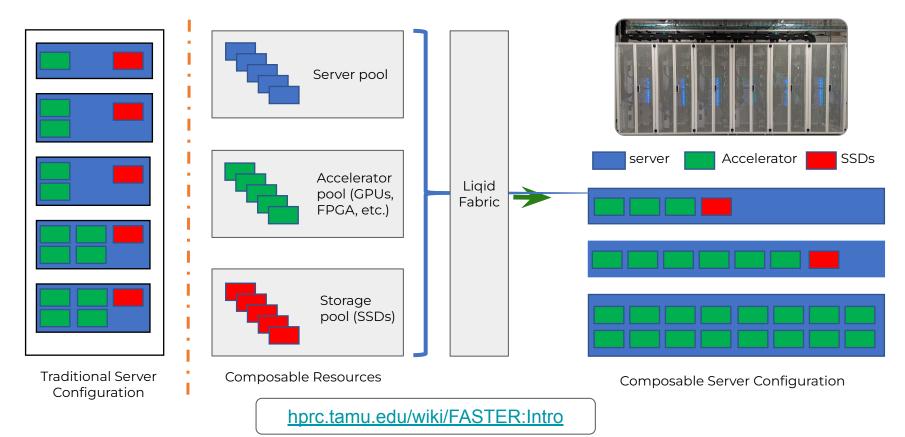
hprc.tamu.edu/wiki/FASTER:Intro

Resources	Quantity
64-core login nodes	4 (3 for TAMU, 1 for ACCESS)
64-core compute nodes (256GB RAM each)	180 (11,520 cores)
Composable GPUs	200 T4 16GB 40 A100 40GB 10 A10 24GB 4 A30 24GB 8 A40 48GB
Interconnect	Mellanox HDR100 InfiniBand (MPI and storage) Liqid PCIe Gen4 (GPU composability)
Global Disk	5PB DDN Lustre appliances



FASTER (Fostering Accelerated Sciences Transformation Education and Research) is a 180-node Intel cluster from Dell featuring the Intel Ice Lake processor.

# Composability at the Hardware Level



# ACES - Accelerating Computing for Emerging Sciences (Phase I)

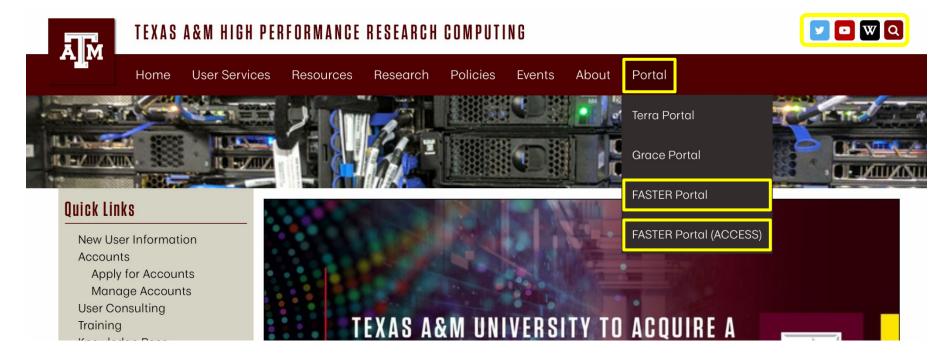


Component	Quantity	Description
<u>Graphcore IPU</u>	16	16 Colossus GC200 IPUs and dual AMD Rome CPU server on a 100 GbE RoCE fabric
Intel FPGA PAC D5005	2	FPGA SOC with Intel Stratix 10 SX FPGAs, 64 bit quad-core Arm Cortex-A53 processors, and 32GB DDR4
Intel Optane SSDs	8	3 TB of Intel Optane SSDs addressable as memory using MemVerge Memory Machine.

ACES Phase I components are available through <u>FASTER</u>

# Accessing the HPRC Portal

• HPRC webpage: <u>hprc.tamu.edu</u>, Portal dropdown menu



# Accessing FASTER via the HPRC Portal (TAMU)

Log-in using your TAMU NetID credentials.

Central Authentic		Activate Your NetID
	LOG IN	
	Current Users	
	NetID or Email Address	
	Password	
	Next Forgot your password?	
	New Student or Employee? Activate your NetID	

# Accessing FASTER via the HPRC Portal (ACCESS)

ACCESS	Powered By		
Consent to Attribute Release TAMUFASTER ACCESS 000 requests access to the following information. If you do not approve this request, do not proceed . Your Clogon user identifier . Your ana . Your sensane and affiliation from your identify provider . Your username and affiliation from your identify provider . Clogon User Section 1 Constrained in the section 1 Constrained in the grass pairs Your Clogon User Section 1 Constrained in the grass pairs Your Clogon User Section 1 Constrained in the section 1 Constrained in the grass pairs Your Section 1 Constrained in the section 1 Constr		Login to CILogon ACCESS Username ACCESS Password COON'L Remember Login Login	Cicloogian Cicloogian Cicloon Access to CyberInfrastructure (Ci). If you had an XSEDE account, please enter your XSEDE username and password for ACCESS logia Pagister for an ACCESS Account Pagister for an ACCESS Account Pagister for an ACCESS Account
Know <u>your responsibilities</u> rusing the Cit.ogon Senice See achigoridedgement of support for this site		Click Here for Assistance	

Select the Identity Provider appropriate

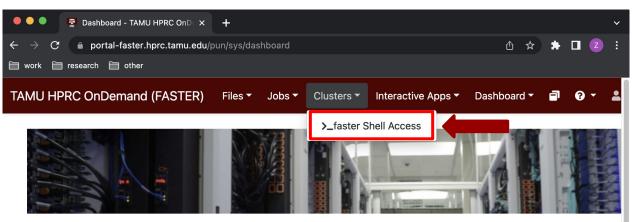
for your account.

Select an Identity Provider

ACCESS CI (XSEDE) \* 0

#### Log-in using your ACCESS credentials.

# Shell Access - I



OnDemand provides an integrated, single access point for all of your HPC resources.

#### Message of the Day

#### **IMPORTANT POLICY INFORMATION**

- Unauthorized use of HPRC resources is prohibited and subject to criminal prosecution.
- Use of HPRC resources in violation of United States export control laws and regulations is prohibited. Current HPRC staff members are US citizens and legal residents.
- Sharing HPRC account and password information is in violation of State Law. Any shared accounts will be DISABLED.
- · Authorized users must also adhere to ALL policies at: https://hprc.tamu.edu/policies

# Shell Access - II

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Website:       https://hprc.tamu.edu                 Consulting:       help@hprc.tamu.edu (preferred) or (979) 845-0219         FASTER Documentation:https://hprc.tamu.edu/wiki/FASTER         Grace Documentation: https://hprc.tamu.edu/wiki/Grace         YouTube Channel:       https://www.youtube.com/texasamhprc				
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* === IMPORTANT POLICY INFORMATION === *				
<ul> <li>* - Unauthorized use of HPRC resources is prohibited and subject to</li> <li>* criminal prosecution.</li> </ul>				
* - Use of HPRC resources in violation of United States export control *				
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<ul> <li>* US citizens and legal residents.</li> <li>* - Sharing HPRC account and password information is in violation of *</li> </ul>				
<ul> <li>* - Sharing HPRC account and password information is in violation of</li> <li>* Texas State Law. Any shared accounts will be DISABLED.</li> </ul>				
* - Authorized users must also adhere to ALL policies at: *				
<pre>* https://hprc.tamu.edu/policies/ *</pre>				
***************************************				
<pre>!! WARNING: THERE ARE ONLY NIGHTLY BACKUPS OF USER HOME DIRECTORIES. !!</pre>				
Please restrict usage to <u>8 CORES</u> across ALL login nodes. Users found in violation of this policy will be <u>SUSPENDED</u> .				
To see these messages again, run the <u>motd</u> command. Your current disk quotas are:				
Disk Disk Usage Limit File Usage Limit /home/happidence1 56K 10.0G 26 10000 /scratch/user/happidence1 631.0G 2.0T 450644 1000000 * Quota increase for /scratch/user/happidence1 will expire on May 21, 2023 /scratch/group/benchmark_prj 325.1G 5.0T 1333878 5000000 /scratch/group/hprc 3.9T 10.0T 615489 1000000 * Quota increase for /scratch/group/hprc will expire on Dec 31, 2026 Type 'showquota' to view these quotas again. (base) [happidence1@faster2 ~]\$				

# Commands to copy the materials

• Navigate to your personal scratch directory

\$cd \$SCRATCH

• Files for this course are located at

/scratch/training/ai\_tech\_labs

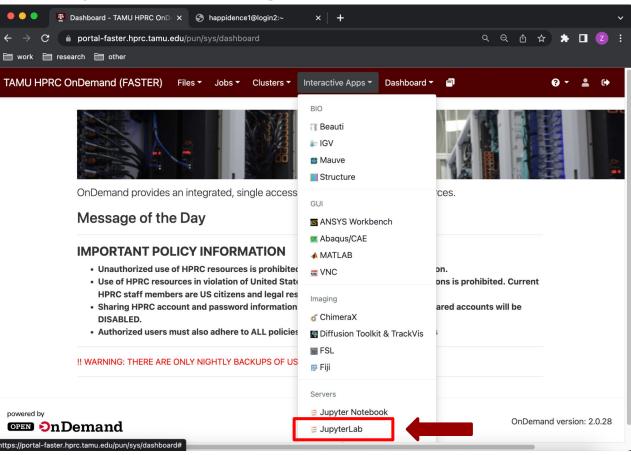
Make a copy in your personal scratch directory

\$ cp -r /scratch/training/ai\_tech\_labs \$SCRATCH

• Enter this directory (your local copy)

\$cd ai\_tech\_labs

# Go to JupyterLab Page

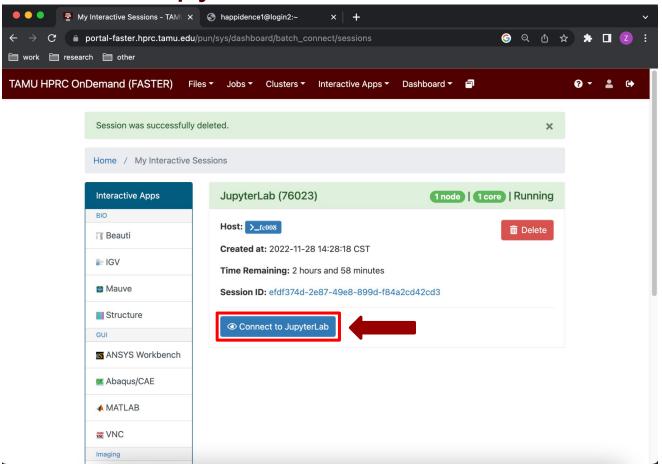


# JupyterLab Page

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Home / My Inte	eractive Sessions / JupyterLab	
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Beauti	cluster. Module	
∎≓ IGV	Python/3.8.2	
Mauve	Anaconda3 uses Python3	
Structure	Optional Environment to be activated	
GUI		Number of hours:
S ANSYS Workb	ench Enter the name of the environment to be activated. (Optional)	
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📣 MATLAB		
🐷 VNC	Leave blank to use the default environment for the selected Module.	Node type: ANY
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of ChimeraX	built with one of the Anaconda or Python modules listed in the Module option above. See instructions.	

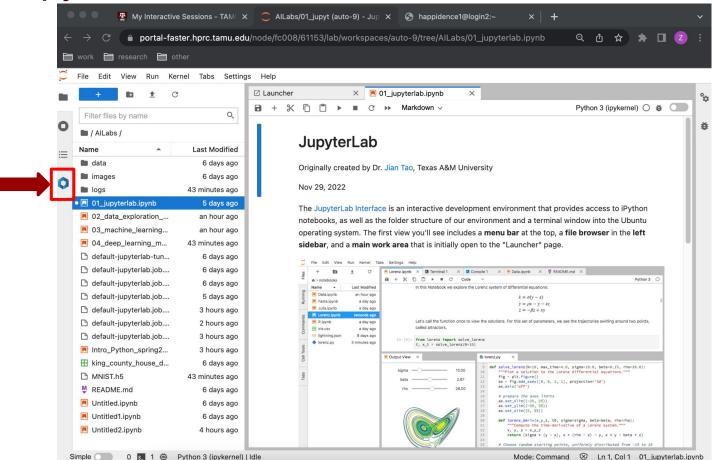
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### Connect to JupyterLab

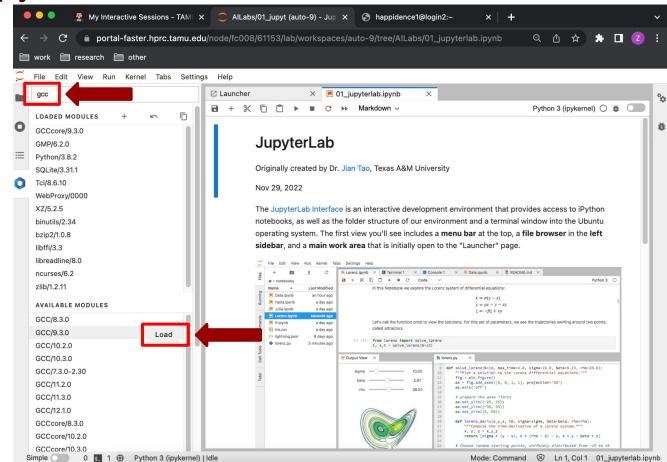


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### JupyterLab Lmod Extension



### JupyterLab Lmod Extension



# **Exercise: Load Required Modules**

- GCC/9.3.0
- OpenMPI/4.0.3
- scikit-learn/0.23.1-Python-3.8.2
- TensorFlow/2.3.1-Python-3.8.2

Note: numpy and matplotlib have already been in the

Scipy-bundle/2020.03-Python-3.8.2 module.

### Test loaded modules

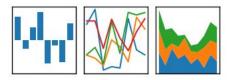
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# Lab II. Data Exploration



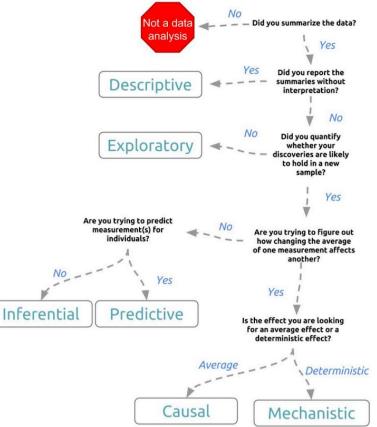




# **Types of Data Science Problems**

- **Descriptive** (summaries, e.g., census)
- Exploratory (search for unknowns, e.g., four-planet solar system)
- Inferential (find correlations, e.g., many social studies)
- Predictive (make predictions, e.g., Face ID, Echo, Siri)
- **Causal** (explore causation, e.g., smoking versus lung cancer)
- Mechanistic (determine governing principles,

e.g., experimental science)



Credit: Jeff Leek - The Elements of Data Analytic Style

# **Data Structures**

Pandas has two data structures that are descriptive and

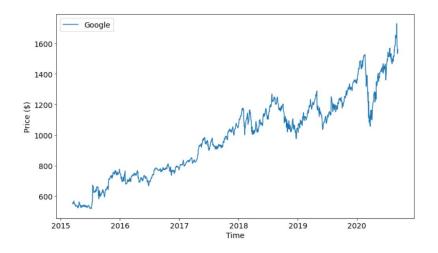
optimized for data with different dimensions.

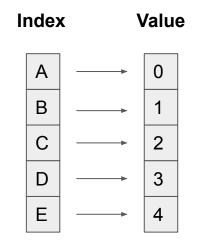
- Series: 1D labeled array
- **DataFrame:** General 2D labeled, size-mutable tabular

structure with potentially heterogeneously-typed columns

# **Series in pandas**

- One-dimensional labeled array
- Capable of holding any data type (integers, strings, floating point numbers, etc.)
- Example: time-series stock price data

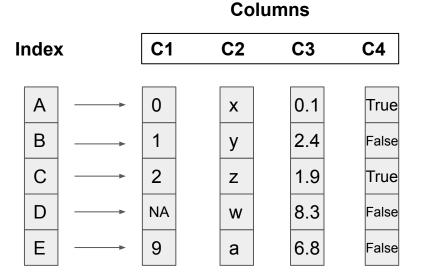




# **DataFrame in pandas**

- Primary Pandas data structure
- A dict-like container for Series objects
- Two-dimensional size-mutable
- Heterogeneous tabular data structure

A	В	с	D	E	F	G	Н
id	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors
7129300520	20141013T0	221900	3	1	1180	5650	1
6414100192	20141209T0	538000	3	2.25	2570	7242	2
5631500400	20150225T0	180000	2	1	770	10000	1
2487200875	20141209T0	604000	4	3	1960	5000	1
1954400510	20150218T0	510000	3	2	1680	8080	1
7237550310	20140512T0	1.23E+06	4	4.5	5420	101930	1
1321400060	20140627T0	257500	3	2.25	1715	6819	2
2008000270	20150115T0	291850	3	1.5	1060	9711	1
2414600126	20150415T0	229500	3	1	1780	7470	1



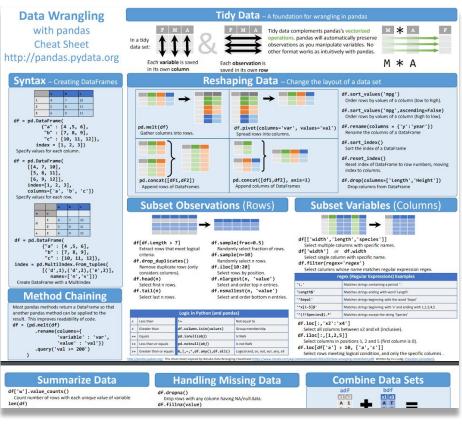
# **Pandas Learning Objectives**

#### After this lesson, you will know how to:

- Create a DataFrame
- Drop Entries
- Index, Select, and Filter data
- Sort data
- Input and Output



# **Pandas Cheat Sheet**



#### https://pandas.pydata.org/Pandas\_Cheat\_Sheet.pdf

# **Key Plotting Concepts in Matplotlib**

#### Matplotlib: Figure

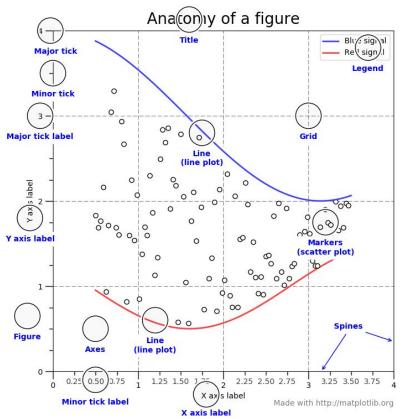
Figure is the object that keeps the whole image output. Adjustable parameters include:

- 1. Image size (set\_size\_inches())
- 2. Whether to use tight\_layout (set\_tight\_layout())

#### Matplotlib: Axes

Axes object represents the pair of axis that contain a single plot (x-axis and y-axis). The Axes object also has more adjustable parameters:

- 1. The plot frame (set\_frame\_on() or set\_frame\_off())
- 2. X-axis and Y-axis limits (set\_xlim() and set\_ylim())
- X-axis and Y-axis Labels (set\_xlabel() and set\_ylabel())
- 4. The plot title (set\_title())



(Credit: matplotlib.org)

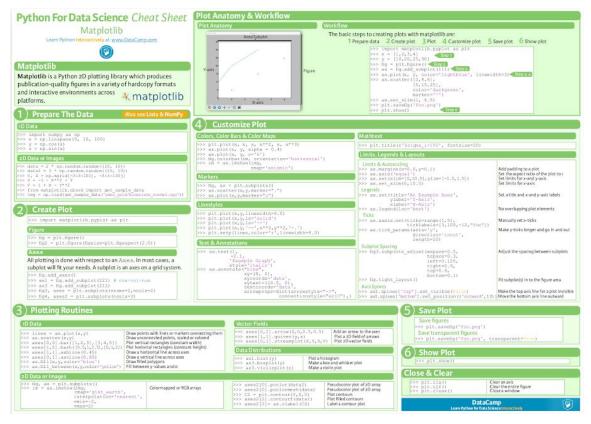
# **Matplotlib Learning Objectives**

After this lesson, you will know how to:

- Scatter plot and Line plot
- Subplots
- Color map
- Contour figures
- 3D figures
  - Surface plots
  - Wire-frame plot
  - Contour plots with projections

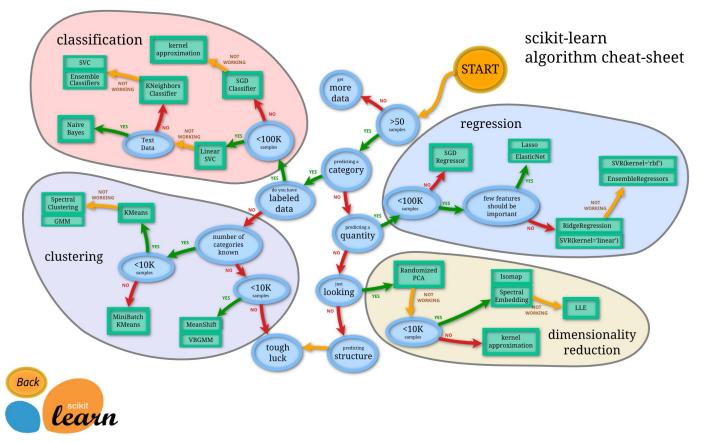


# **Matplotlib Cheat Sheet**



https://s3.amazonaws.com/assets.datacamp.com/blog\_assets/Python\_Matplotlib\_Cheat\_Sheet.pdf

# Lab III. Machine Learning



# **Main Features of scikit-learn**



Classification	Classification Regression		Dimension Reduction	Model Selection	Preprocessing
Identifying category of an object	Predicting a attribute for an object	Grouping similar objects into sets	Reducing the number of dimensions	Selecting models with parameter search	Preprocessing data to prepare for modeling
Applications: Spam detection, image recognition. Algorithms: SVM, nearest neighbors, random forest, and more	<b>Applications</b> : Drug response, Stock prices. <b>Algorithms</b> : SVR, nearest neighbors, random forest, and more	<b>Applications:</b> Customer segmentation, Grouping experiment outcomes <b>Algorithms:</b> k-Means, spectral clustering, mean-shift, and more	<b>Applications:</b> Visualization, Increased efficiency <b>Algorithms:</b> k-Means, feature selection, non-negative matrix factorization, and more	<b>Applications:</b> Improved accuracy via parameter tuning <b>Algorithms:</b> grid search, cross validation, metrics, and more	<b>Applications:</b> Transforming input data such as text for use with machine learning algorithms. <b>Algorithms:</b> preprocessing, feature extraction, and more
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Credit: icons are from The Noun Project under Creative Commons Licenses

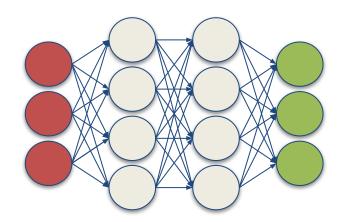
# Lab IV. Deep Learning

**Deep Learning** by Ian Goodfellow, Yoshua Bengio, and Aaron Courville <u>http://www.deeplearningbook.org/</u>

#### Animation of Neutron Networks

by Grant Sanderson https://www.3blue1brown.com/

*Visualization of CNN* by Adam Harley <u>https://adamharley.com/nn\_vis/cnn/3d.html</u>

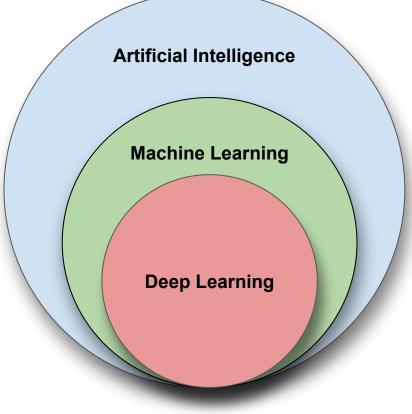






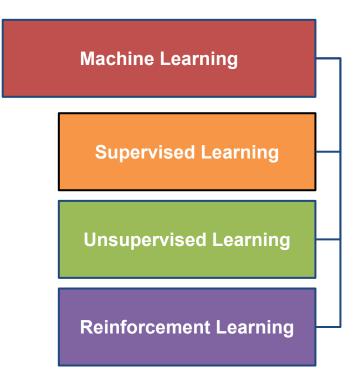
# Relationship of AI, ML, and DL

- Artificial Intelligence (AI) is anything about man-made intelligence exhibited by machines.
- Machine Learning (ML) is an approach to achieve AI.
- Deep Learning (DL) is one technique to implement ML.

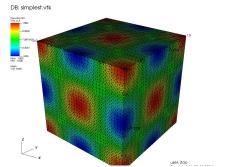


# **Types of ML Algorithms**

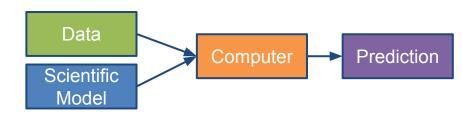
- Supervised Learning
  - trained with labeled data; including regression and classification problems
- Unsupervised Learning
  - trained with unlabeled data; clustering and association rule learning problems.
- Reinforcement Learning
  - no training data; stochastic
     Markov decision process; robotics and business strategy planning.



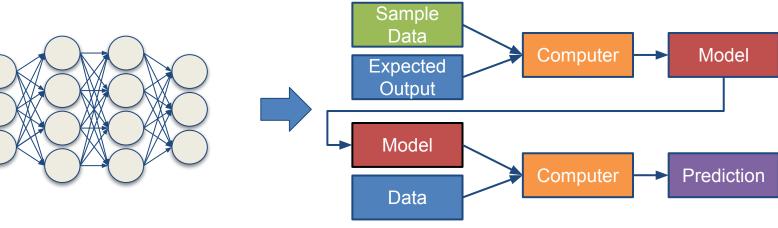
# **Machine Learning**



#### **Traditional Modeling**



#### Machine Learning (Supervised Learning)



# **Inputs and Outputs**

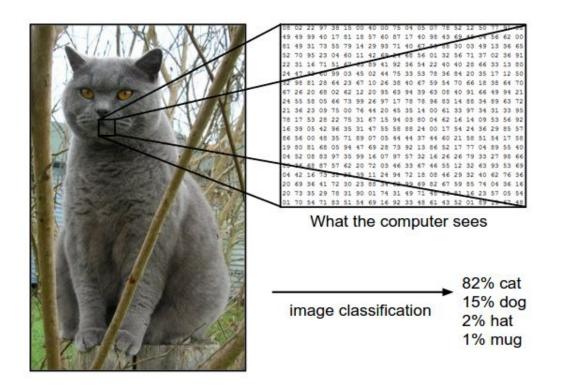
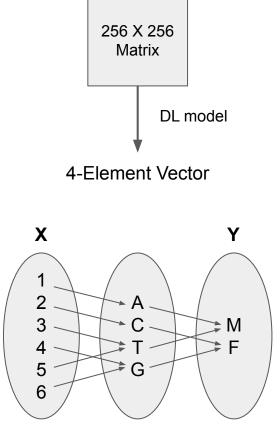
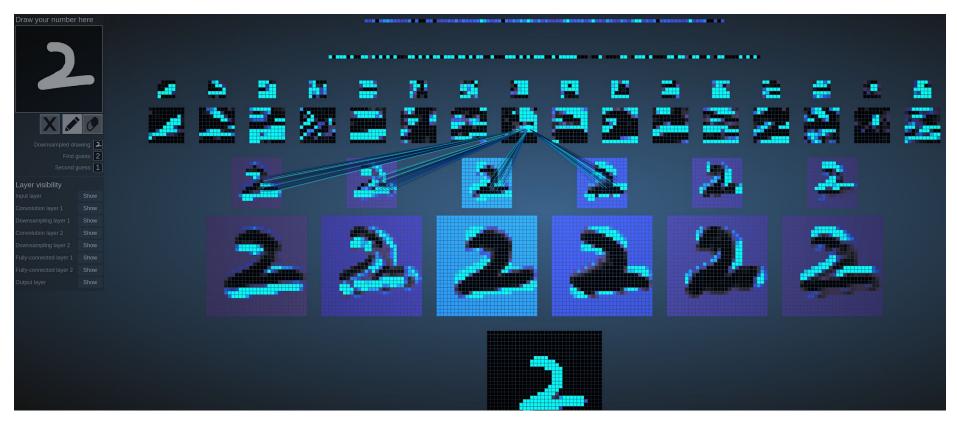


Image from the Stanford CS231 Course



With deep learning, we are searching for a **surjective** (or **onto**) function **f** from a set **X** to a set **Y**.

### **MNIST - CNN Visualization**



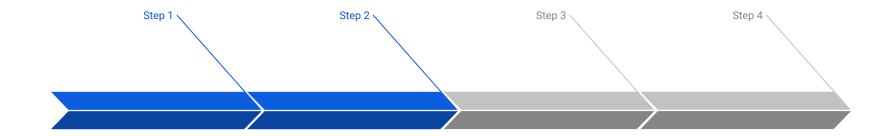
(Image Credit: https://adamharley.com/nn\_vis/cnn/3d.html)

# **CNN Explainer**



(Image Credit: https://poloclub.github.io/cnn-explainer/)

# **Machine Learning Workflow with Keras**



#### Prepare Train Data

The preprocessed data set needs to be shuffled and splitted into training and testing data.

#### **Define Model**

A model could be defined with Keras Sequential model for a linear stack of layers or Keras functional API for complex network.

#### **Training Configuration**

The configuration of the training process requires the specification of an optimizer, a loss function, and a list of metrics.

#### **Train Model**

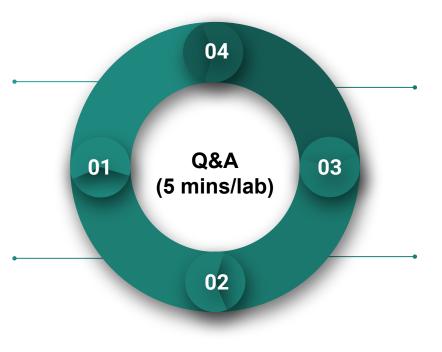
The training begins by calling the fit function. The number of epochs and batch size need to be set. The measurement metrics need to be evaluated.

# AI Tech Labs

Lab I.

(30 mins)

JupyterLab



#### Lab IV. Deep Learning (30 minutes)

We will learn how to use Keras to build and train a simple image classification model with deep neural network (DNN).

#### Lab III Machine Learning (30 minutes)

We will learn to use scikit-learn library for linear regression and classification applications.

Lmode extension and run JupyterLab on HPRC portal. Lab II.

We will load required

modules with Jupyter

#### Data Exploration (30 mins)

We will go through some examples with two popular Python libraries: Pandas and Matplotlib for data exploration.

Figure 1. Structure of the AI Technology Labs.