

Neural Network

with **TensorFlow**



```
import tensorflow as tf
```



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@Future_Rainbow

```
def filterStudies(studies, filterByOrg : bool, filterByStatus : bool) : List[Study] {
  val filteredStudies = studies.filter(study => {
    (filterByOrg == false || study.lead_organization == filterByOrg) &&
    (filterByStatus == false || study.status == filterByStatus)
  })
  filteredStudies
}
```

Teaser

What we are going to learn?

- Introduction to **tensor**
- Preprocess and load data into tensors
- Build a neural network with Keras **Sequential/Functional** API
- Working on a small **regression** example
- Build a **multi-class classification** model
- Improve the model by **transfer learning** (EfficientNetB0)

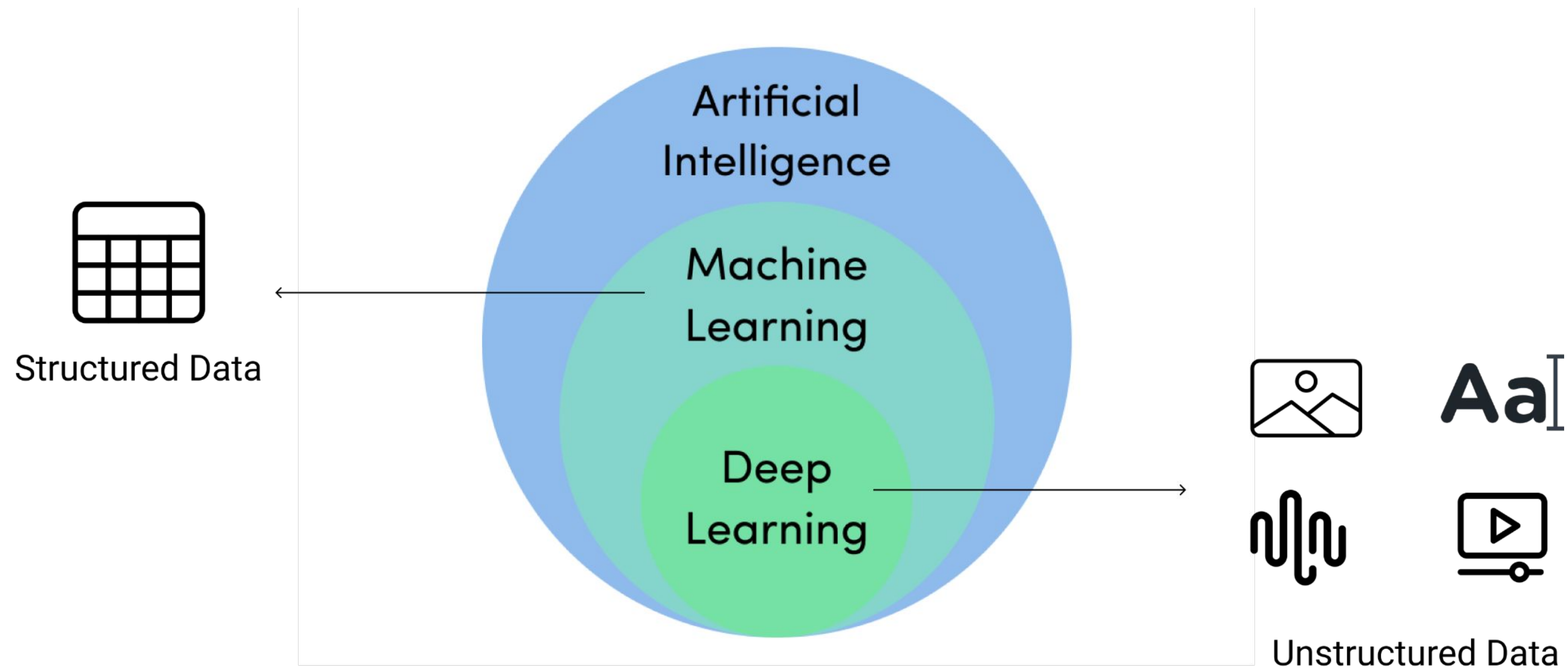


```
tensor = tf.constant(123)
```

```
model = tf.keras.Sequential([  
    tf.keras.layers.Dense(1), # Input layer  
    tf.keras.layers.Dense(10), # Hidden layer  
    tf.keras.layers.Dense(100), # Hidden layer  
    tf.keras.layers.Dense(1), # Output layer  
)
```


Before that

Let's revise something:



Before that

Let's revise something: (commonly used algorithms)

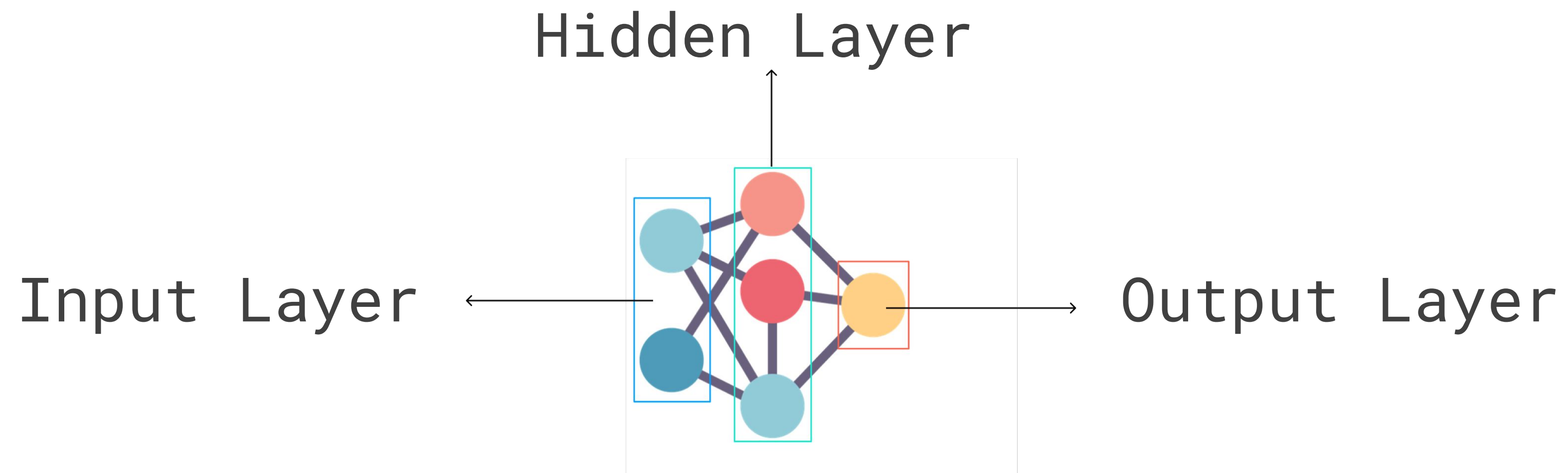
Machine Learning

- Random forest
- Naive bayes
- Nearest neighbour
- SVM
- ...many more

Deep Learning

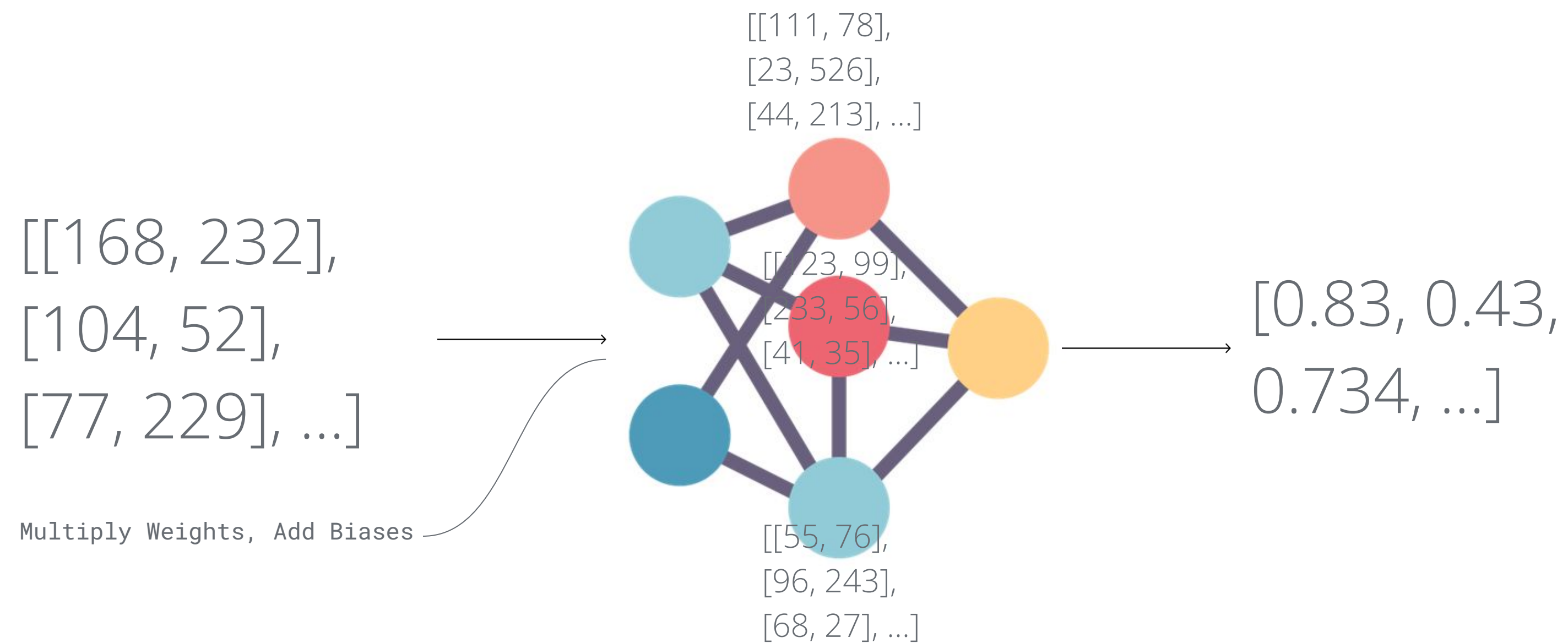
- Neural networks
- Fully connected neural network
- Convolutional neural network
- Recurrent neural network
- ...many more

Breaking down into a Neural Network



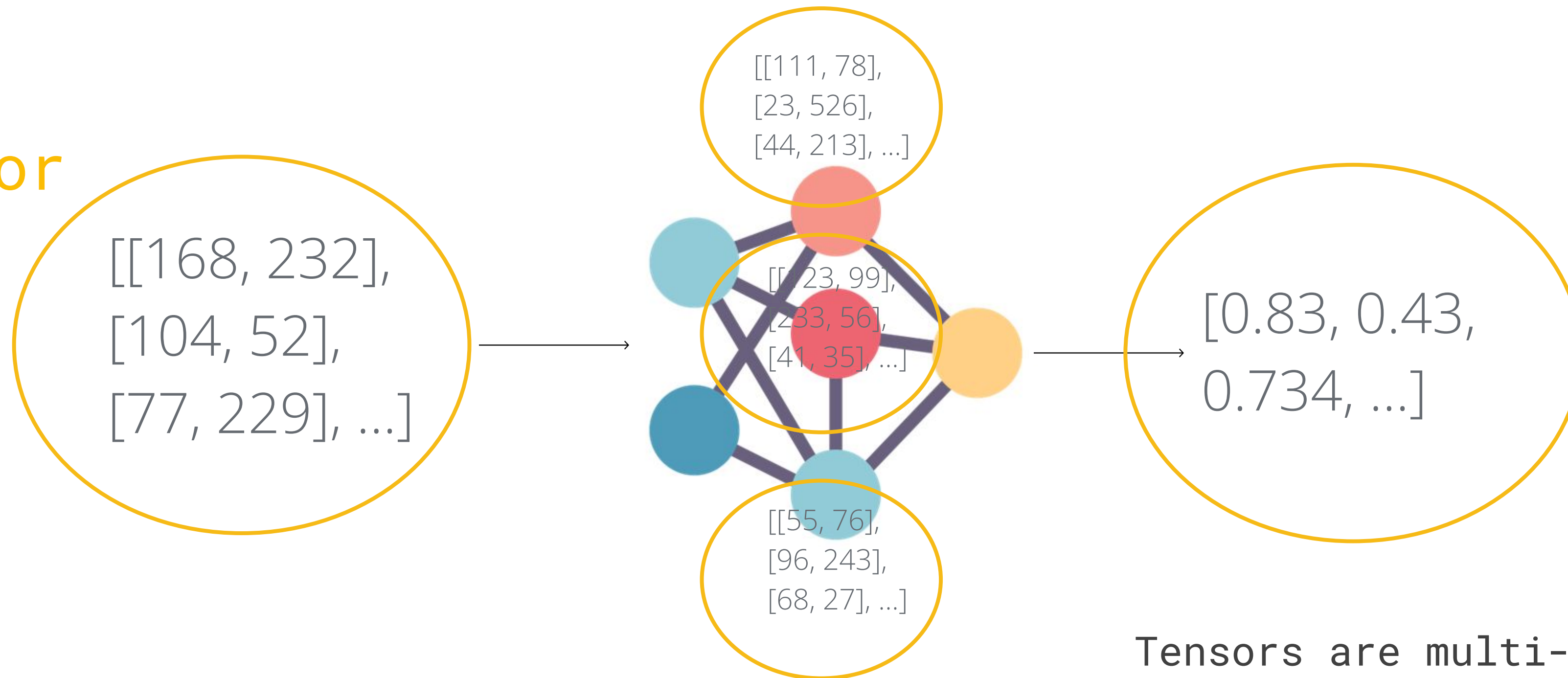
```
model = tf.keras.Sequential([  
    tf.keras.layers.Input(units=2), # Input layer  
    tf.keras.layers.Dense(units=3), # Hidden layer  
    tf.keras.layers.Dense(units=1), # Output layer  
)
```


Breaking down into a Neural Network



Breaking down into a Neural Network

Tensor



Tensors are multi-dimensional arrays with a uniform dtype

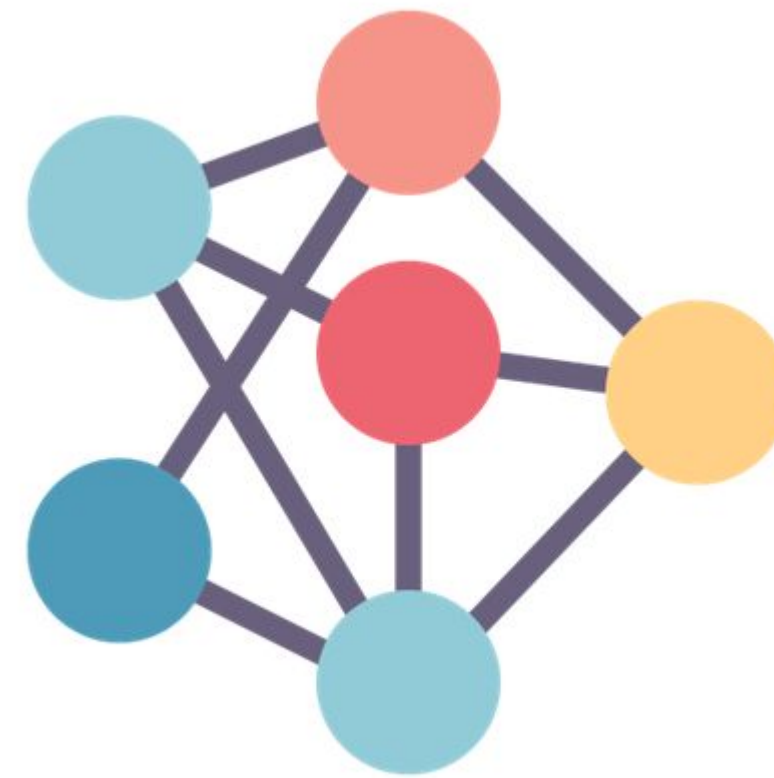
Breaking down into a Neural Network



Image turns into tensors

$[[168, 232],$
 $[104, 52],$
 $[77, 229], \dots]$

Fit

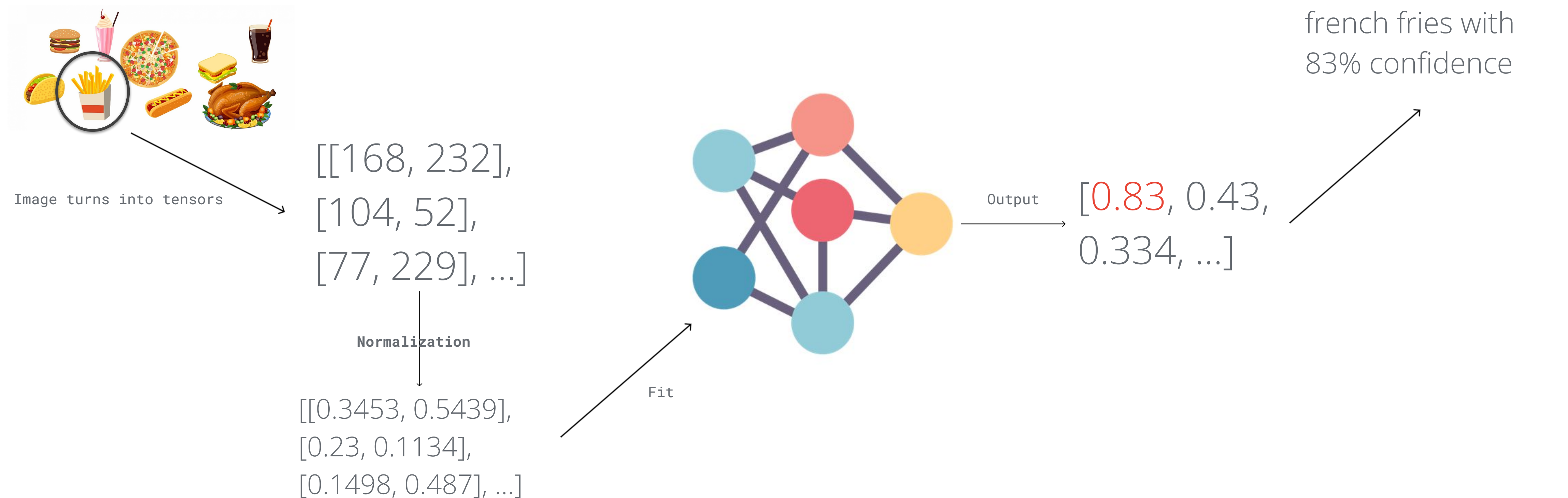


Output

$[0.83, 0.43,$
 $0.334, \dots]$

This image contain
french fries with
83% confidence

Breaking down into a Neural Network



Typical Workflow

for creating a neural network

1. Prepare the data (turn it into tensors)
2. Define the layers and units
3. Compile the model
4. Fit the data
5. Evaluate the model
6. Repeat by experimentation

Typical Workflow

for creating a neural network

- 1. Prepare the data (turn it into tensors)**
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1. Prepare the data (turn it into tensors)

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4. Fit the data

5. Evaluate the model

6. Repeat by experimentation



```
import tensorflow as tf

model = tf.keras.Sequential([
    tf.keras.layers.Input(3), # Input layer
    tf.keras.layers.Dense(10, activation='relu'), # Hidden layer
    tf.keras.layers.Dense(100, activation='relu'), # Hidden layer
    tf.keras.layers.Dense(2, activation='sigmoid') # Output layer
])
```

Typical Workflow

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1. Prepare the data (turn it into tensors)
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- 3. Compile the model**
4. Fit the data
5. Evaluate the model
6. Repeat by experimentation



```
model.compile(  
    loss=tf.keras.losses.binary_crossentropy,  
    optimizer=tf.keras.optimizers.Adam(),  
    metrics=['accuracy']  
)
```

Typical Workflow

for creating a neural network

1. Prepare the data (turn it into tensors)
2. Define the layers and units
3. Compile the model
- 4. Fit the data**
5. Evaluate the model
6. Repeat by experimentation



```
model.fit(  
    X_train,  
    y_train,  
    epochs=5,  
    validation_data=(X_test, y_test)  
)
```


Typical Workflow

for creating a neural network

1. Prepare the data (turn it into tensors)
2. Define the layers and units
3. Compile the model
4. Fit the data
- 5. Evaluate the model**
6. Repeat by experimentation



```
model.evaluate(x_test, y_test)
```

Typical Workflow

for creating a neural network

1. Prepare the data (turn it into tensors)
2. Define the layers and units
3. Compile the model
4. Fit the data
5. Evaluate the model
- 6. Repeat by experimentation**

Example

Regression Problems

Build a neural network that can predict the

housing price based on:

1. No. of bedroom
2. No. of bathroom
3. No. of parking slot



🛏 x 4
🛁 x 4
🚗 x 2

```
# 1. Create a model (specified to your problem)
model = tf.keras.Sequential([
    tf.keras.Input(shape=(3,)),
    tf.keras.layers.Dense(100, activation="relu"),
    tf.keras.layers.Dense(100, activation="relu"),
    tf.keras.layers.Dense(100, activation="relu"),
    tf.keras.layers.Dense(1, activation=None)
])

# 2. Compile the model
model.compile(loss=tf.keras.losses.mae,
              optimizer=tf.keras.optimizers.Adam(lr=0.0001),
              metrics=["mae"])

# 3. Fit the model
model.fit(X_train, y_train, epochs=100)
```

\$ 956,000

Cheatsheet

Regression Problems



🛏 x 4
🛁 x 4
🚗 x 2

Hyperparameter	Typical value
Input layer shape	Same shape as number of features (e.g. 3 for # bedrooms, # bathrooms, # car spaces in housing price prediction)
Hidden layer(s)	Problem specific, minimum = 1, maximum = unlimited
Neurons per hidden layer	Problem specific, generally 10 to 100
Output layer shape	Same shape as desired prediction shape (e.g. 1 for house price)
Hidden activation	Usually ReLU (rectified linear unit)
Output activation	None, ReLU, logistic/tanh
Loss function	MSE (mean square error) or MAE (mean absolute error)/Huber (combination of MAE/MSE) if outliers
Optimizer	SGD (stochastic gradient descent), Adam

Source: Adapted from page 293 of [Hands-On Machine Learning with Scikit-Learn, Keras & TensorFlow Book](#) by Aurélien Géron

```
# 1. Create a model (specified to your problem)
model = tf.keras.Sequential([
    tf.keras.Input(shape=(3,)),
    tf.keras.layers.Dense(100, activation="relu"),
    tf.keras.layers.Dense(100, activation="relu"),
    tf.keras.layers.Dense(100, activation="relu"),
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])

# 2. Compile the model
model.compile(loss=tf.keras.losses.mae,
              optimizer=tf.keras.optimizers.Adam(lr=0.0001),
              metrics=["mae"])

# 3. Fit the model
model.fit(X_train, y_train, epochs=100)
```

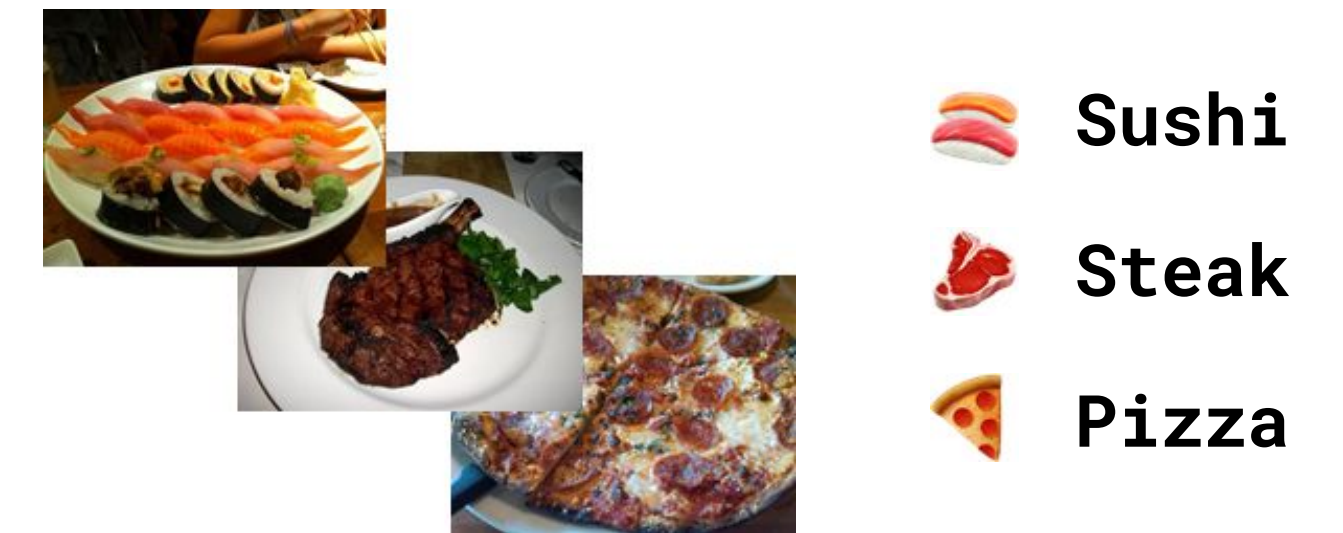
\$ 956,000

Example

Classification Problems

Build a neural network that classifies the images between:

1. Sushi
2. Steak
3. Pizza



```
# 1. Create a model (specified to your problem)
model = tf.keras.Sequential([
    tf.keras.Input(shape=(224, 224, 3)),
    tf.keras.layers.Dense(100, activation="relu"),
    tf.keras.layers.Dense(3, activation="softmax")
])

# 2. Compile the model
model.compile(loss=tf.keras.losses.CategoricalCrossentropy(),
              optimizer=tf.keras.optimizers.Adam(),
              metrics=["accuracy"])

# 3. Fit the model
model.fit(X_train, y_train, epochs=5)

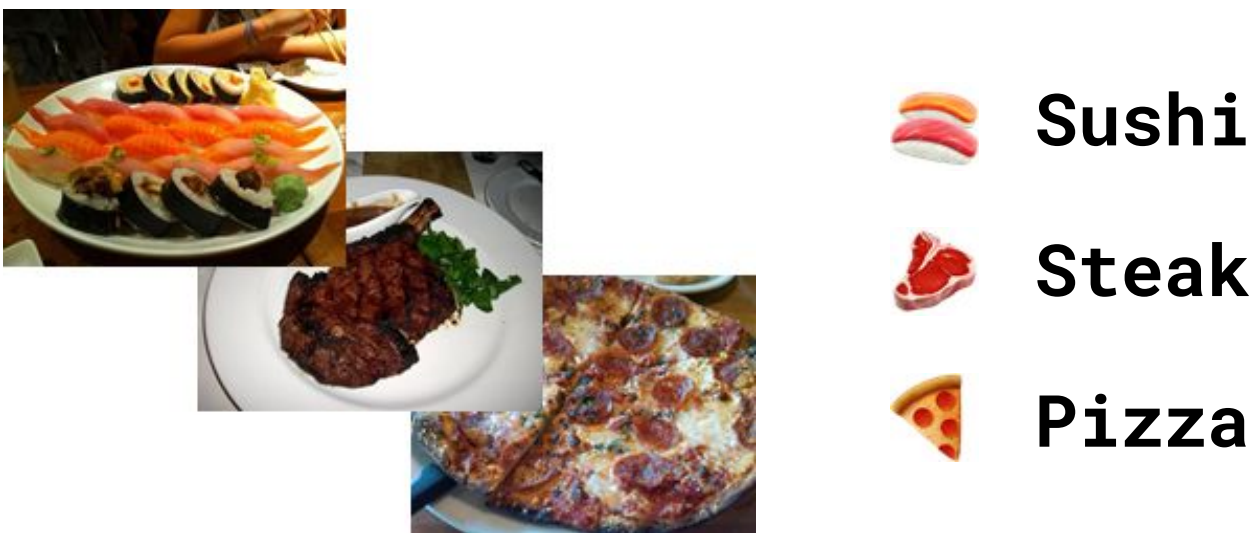
# 4. Evaluate the model
model.evaluate(X_test, y_test)
```

 **Sushi**

`[[0.93, 0.13, 0.236], ...]`

Example

Classification Problems



Hyperparameter	Binary Classification	Multiclass classification
Input layer shape	Same as number of features (e.g. 5 for age, sex, height, weight, smoking status in heart disease prediction)	Same as binary classification
Hidden layer(s)	Problem specific, minimum = 1, maximum = unlimited	Same as binary classification
Neurons per hidden layer	Problem specific, generally 10 to 100	Same as binary classification
Output layer shape	1 (one class or the other)	1 per class (e.g. 3 for food, person or dog photo)
Hidden activation	Usually ReLU (rectified linear unit)	Same as binary classification
Output activation	Sigmoid	Softmax
Loss function	Cross entropy (tf.keras.losses.BinaryCrossentropy in TensorFlow)	Cross entropy (tf.keras.losses.CategoricalCrossentropy in TensorFlow)
Optimizer	SGD (stochastic gradient descent), Adam	Same as binary classification

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```
# 1. Create a model (specified to your problem)
model = tf.keras.Sequential([
    tf.keras.Input(shape=(224, 224, 3)),
    tf.keras.layers.Dense(100, activation="relu"),
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])

# 2. Compile the model
model.compile(loss=tf.keras.losses.CategoricalCrossentropy(),
              optimizer=tf.keras.optimizers.Adam(),
              metrics=["accuracy"])

# 3. Fit the model
model.fit(X_train, y_train, epochs=5)

# 4. Evaluate the model
model.evaluate(X_test, y_test)
```

Sushi

`[[0.93, 0.13, 0.236], ...]`

Cheatsheet

Classification Problems



 x 4
 x 4
 x 2

Hyperparameter	Typical value
Input layer shape	Same shape as number of features (e.g. 3 for # bedrooms, # bathrooms, # car spaces in housing price prediction)
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    tf.keras.layers.Dense(100, activation="relu"),
    tf.keras.layers.Dense(1, activation=None)
])

# 2. Compile the model
model.compile(loss=tf.keras.losses.mae,
              optimizer=tf.keras.optimizers.Adam(lr=0.0001),
              metrics=["mae"])

# 3. Fit the model
model.fit(X_train, y_train, epochs=100)
    
```

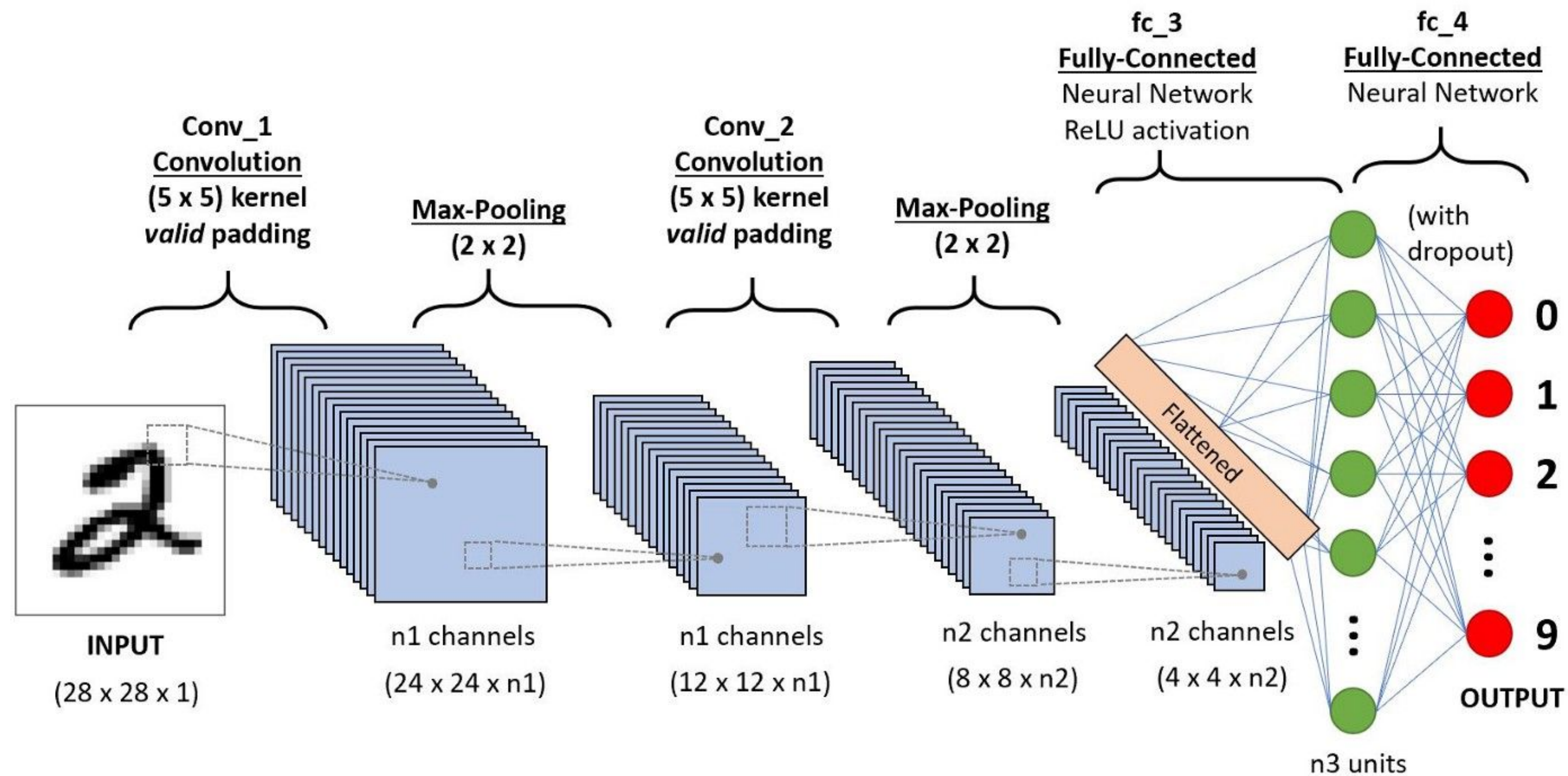
\$ 956,000

Desmos Graph Explanation

Visualize Activation Functions

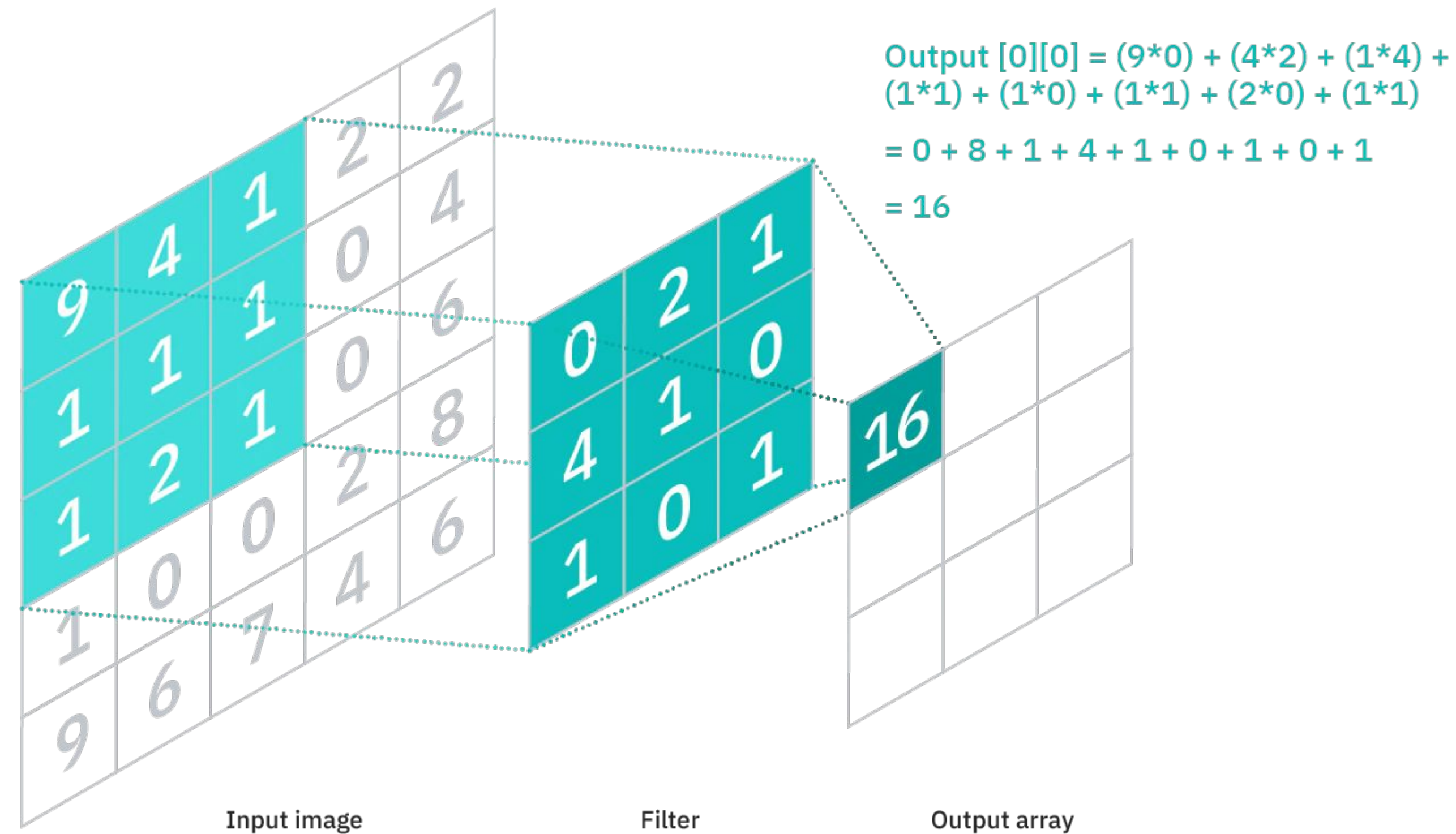
<https://www.desmos.com/calculator/drqqhtb037>

Convolution Neural Network (CNN)



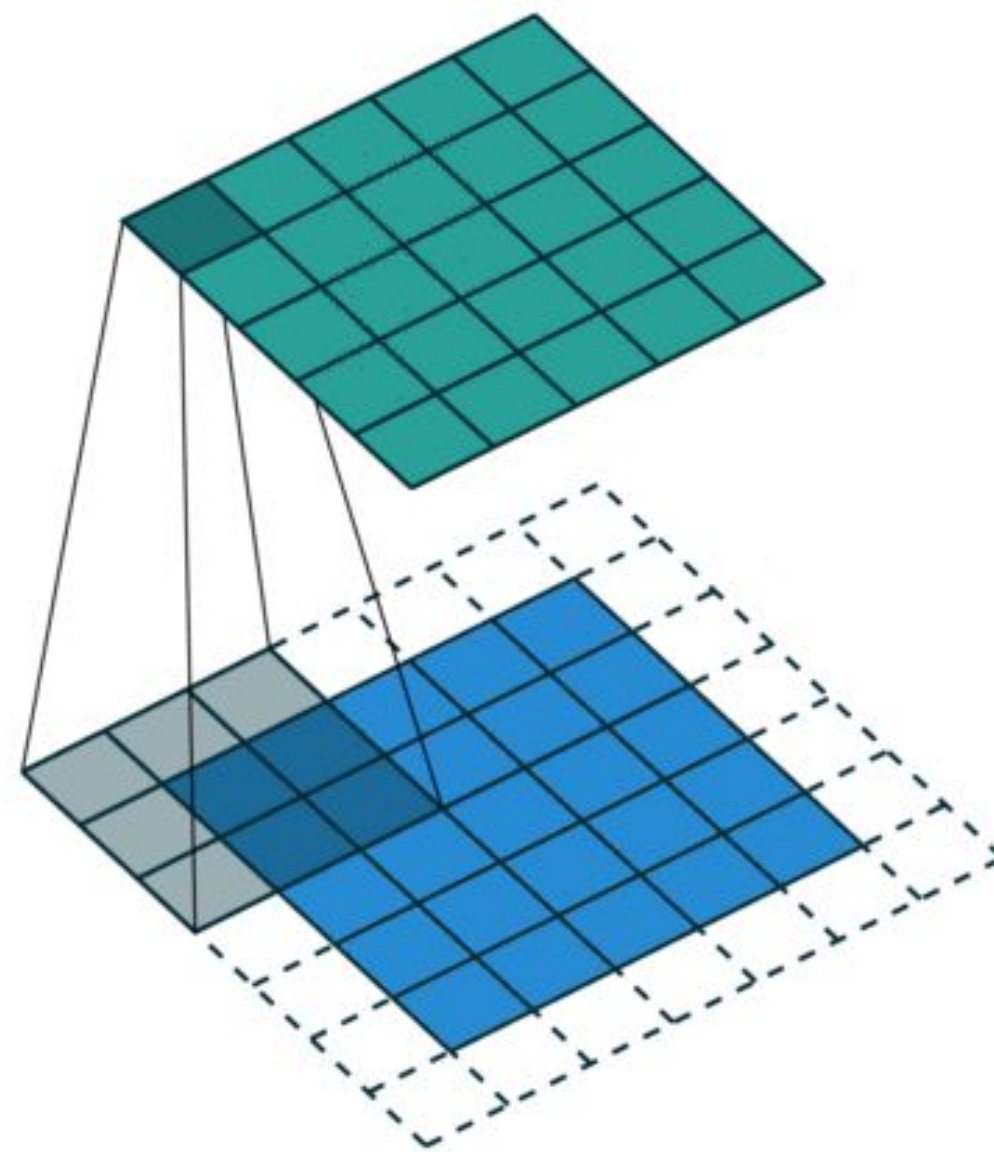
Conv2D Layer

`tf.keras.layers.Conv2D` or `tf.keras.layers.Convolution2D`



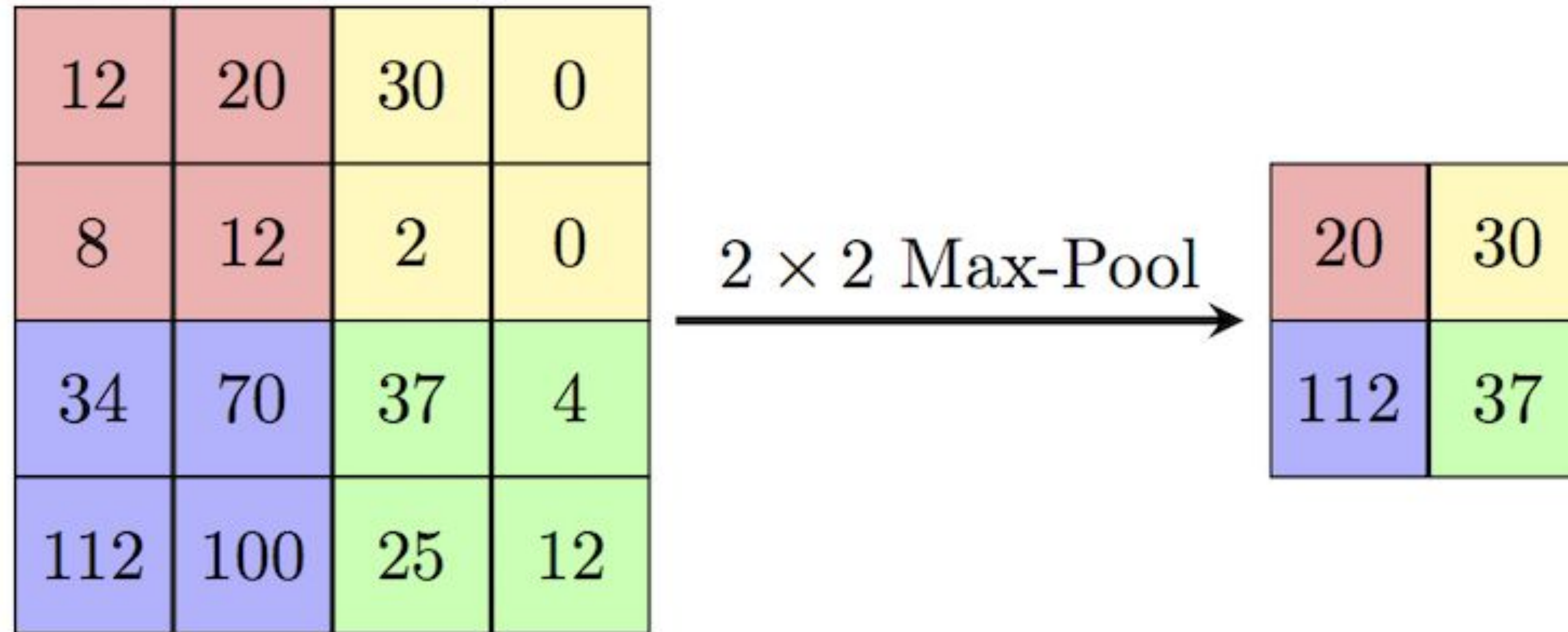
Conv2D Layer

`tf.keras.layers.Conv2D` or `tf.keras.layers.Convolution2D`



MaxPool2D Layer

`tf.keras.layers.MaxPool2D`

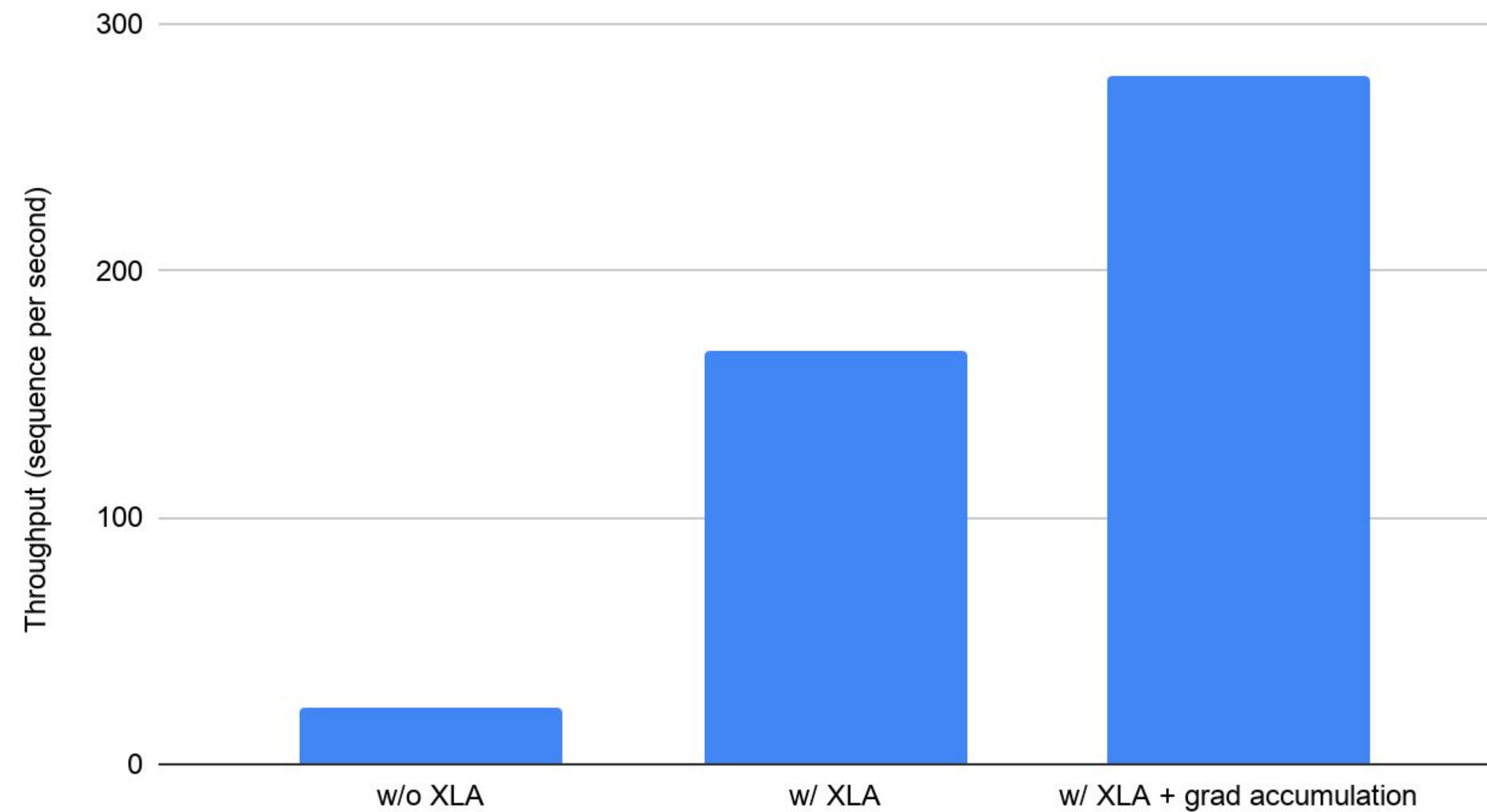


Where to go from here?

Further ML topics to explore

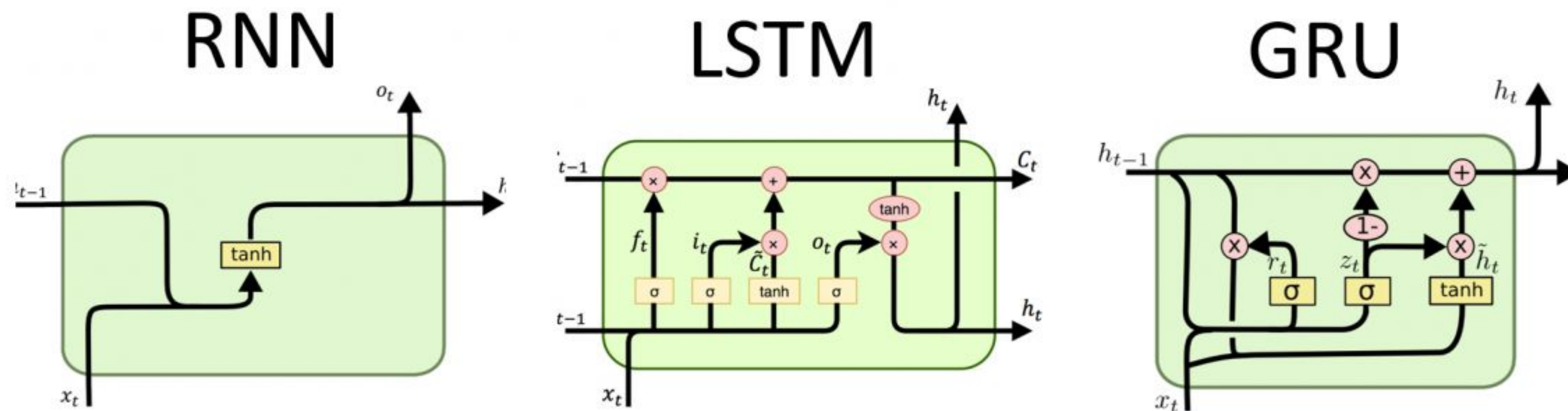
XLA Tensorflow

Accelerated Linear Algebra

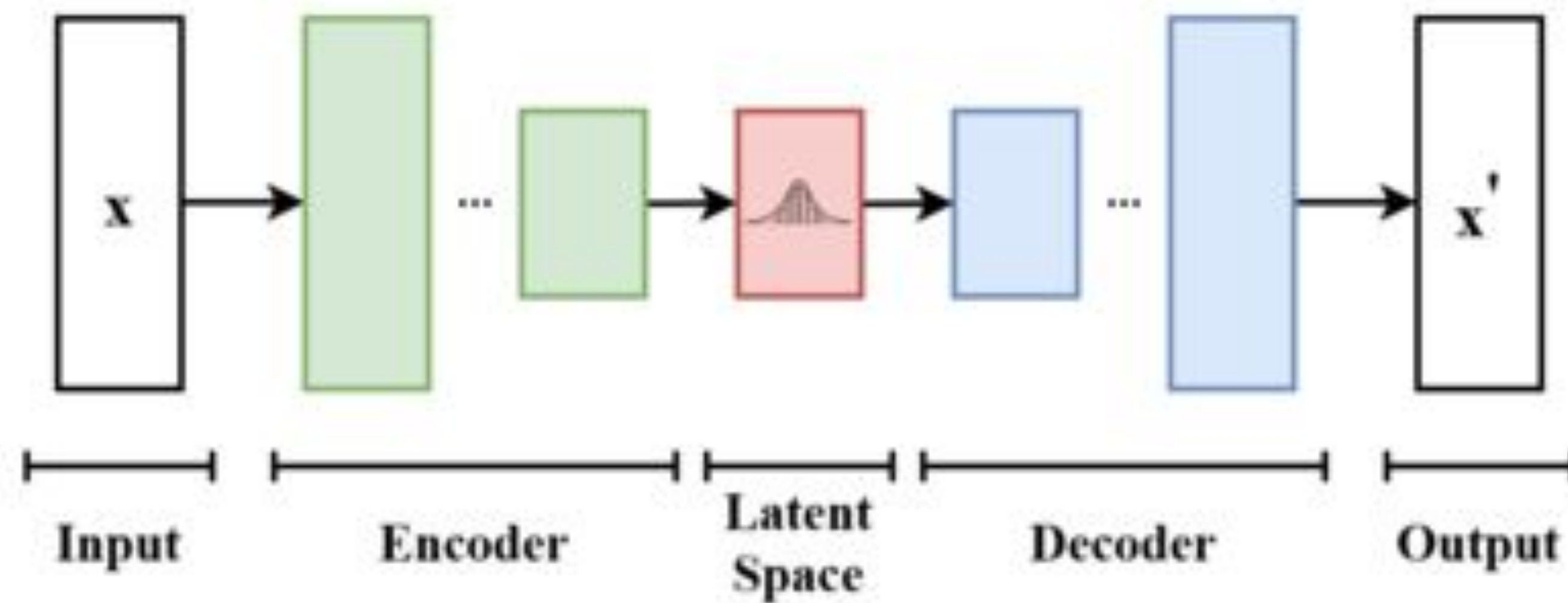


RNN/LSTM/GRU

Different types of neural network architectures

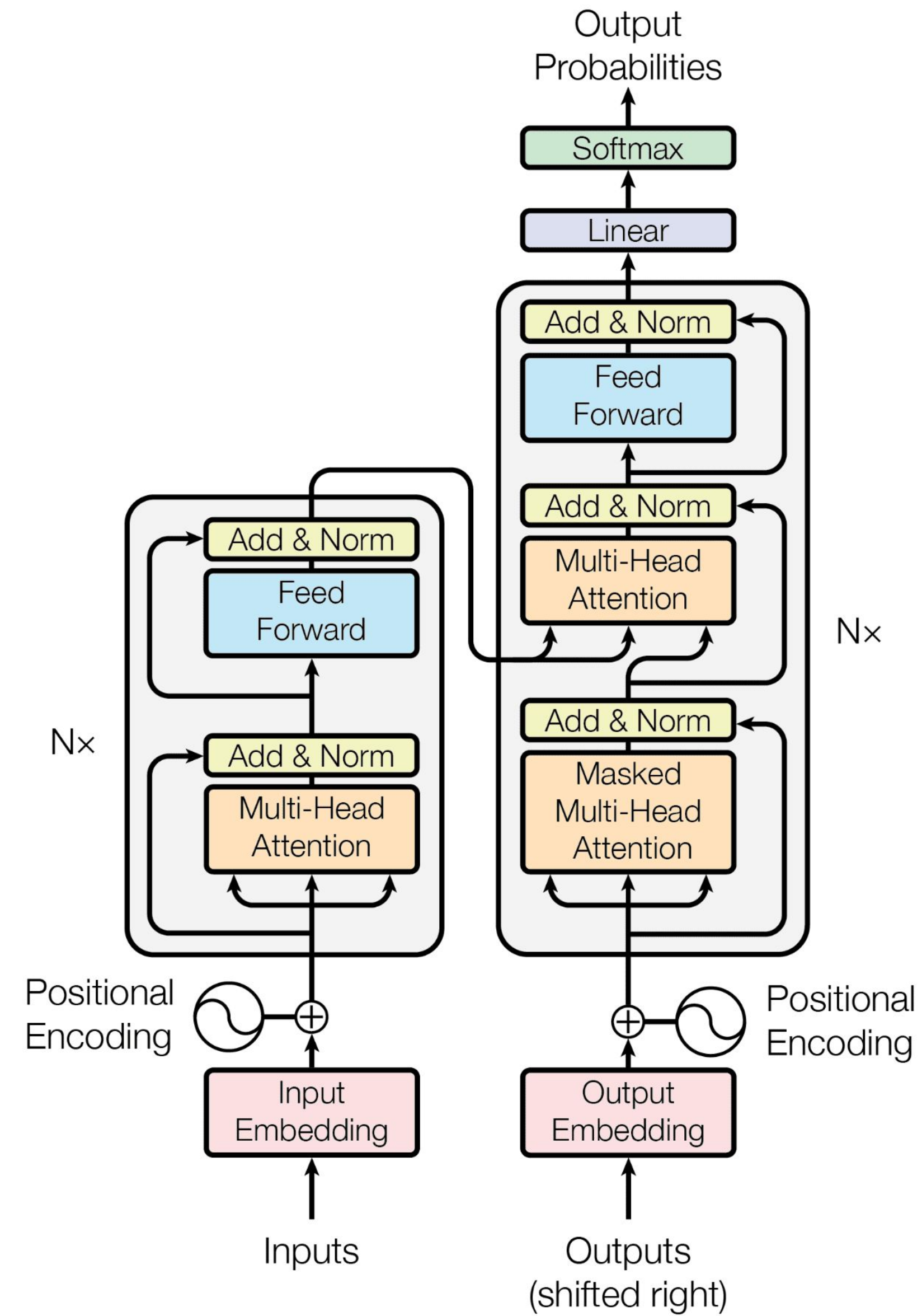


Variational Autoencoder

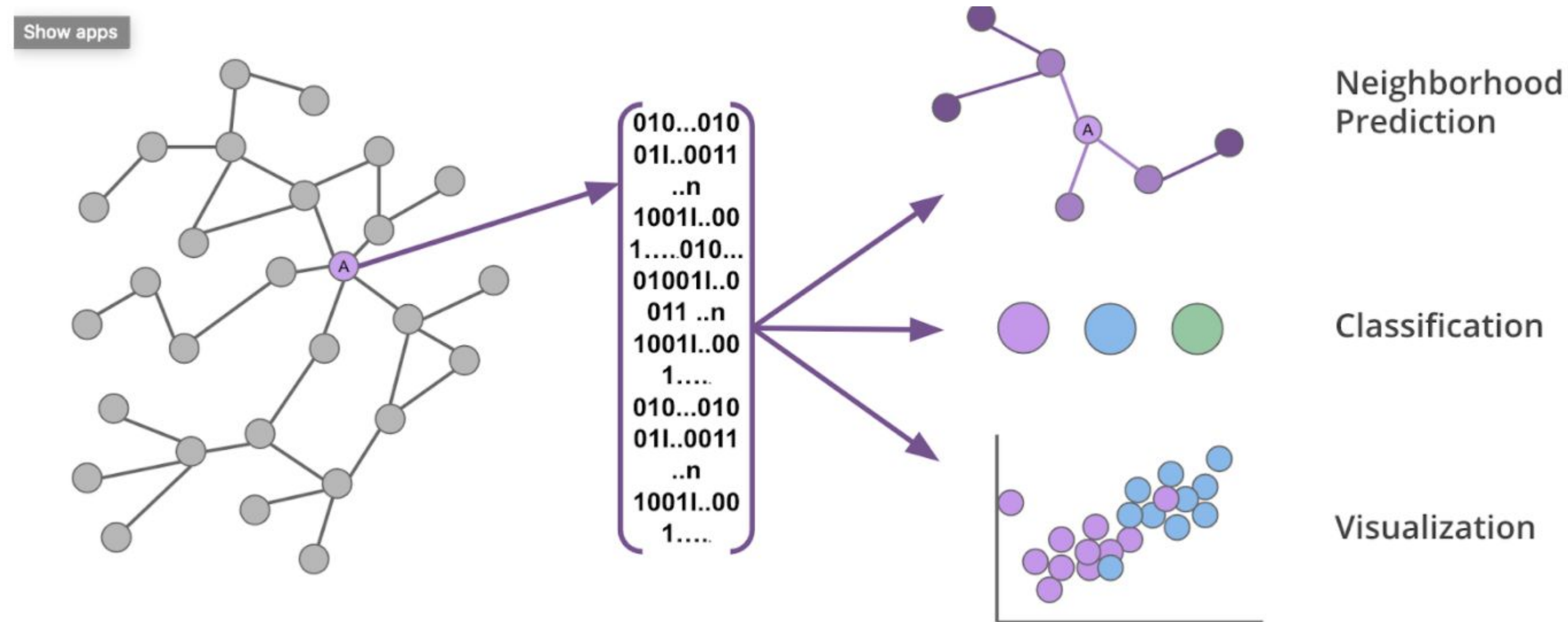


Transformers

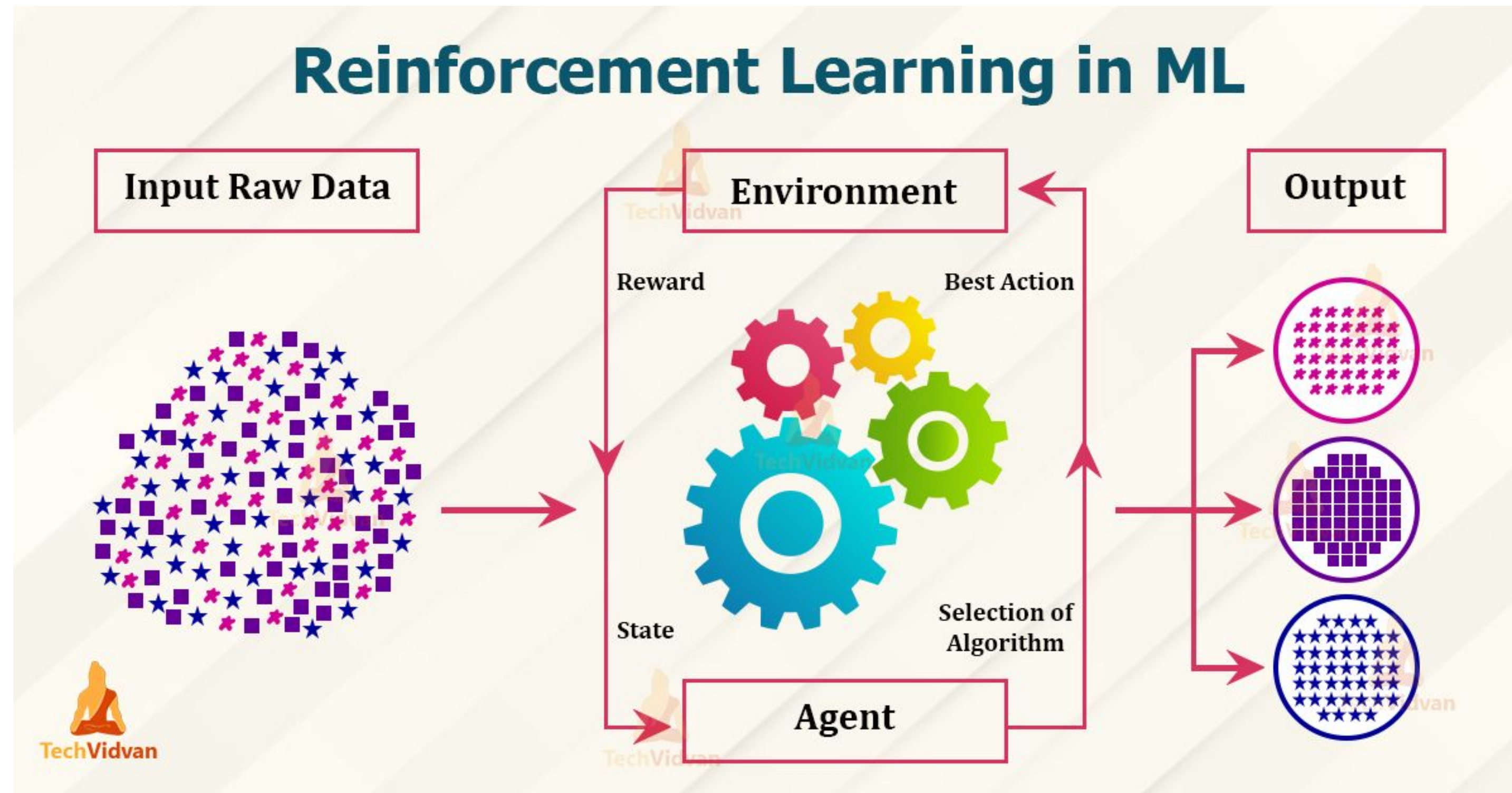
Attention is all you need



Graph Machine Learning

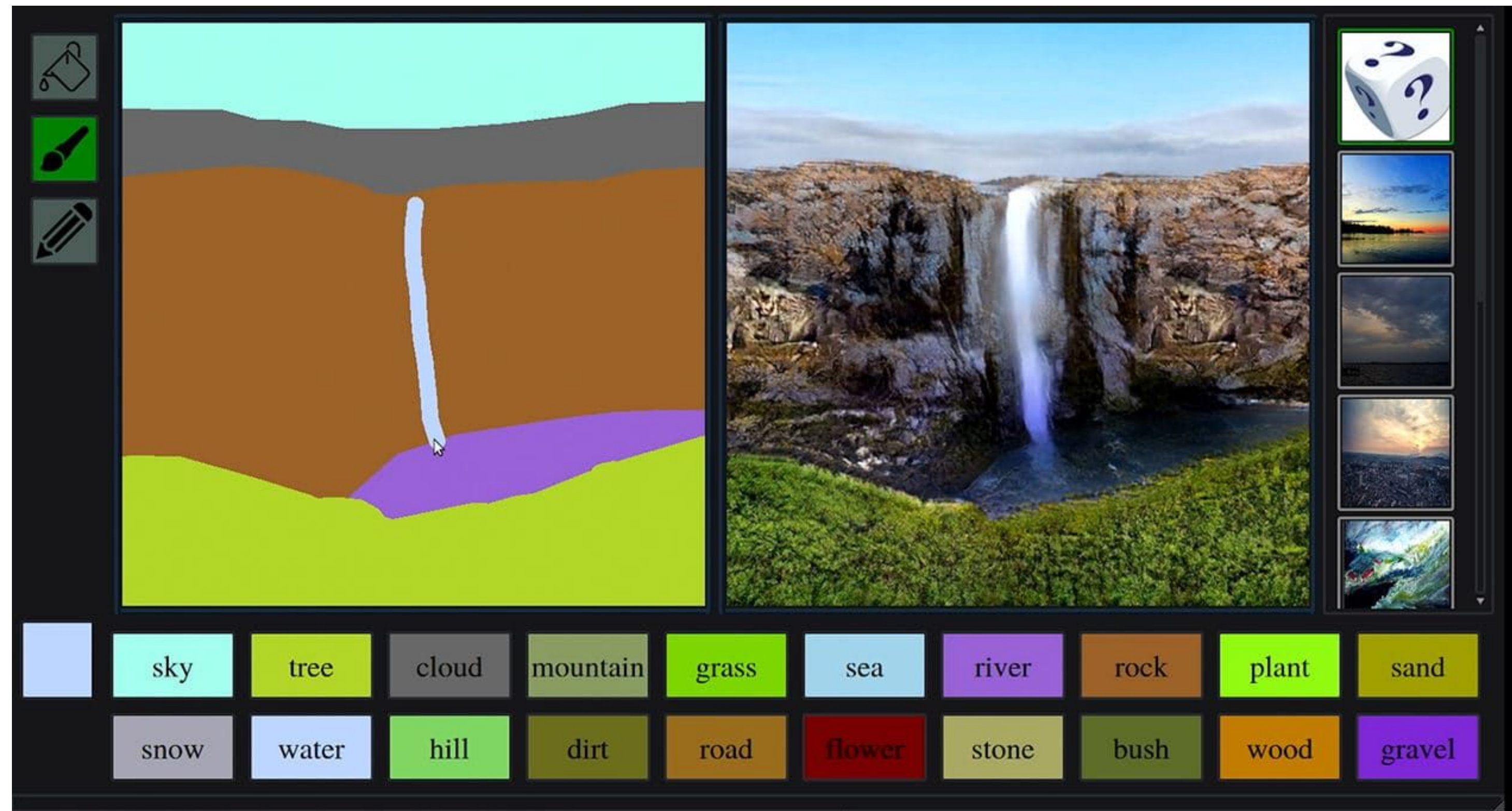


Reinforcement Learning



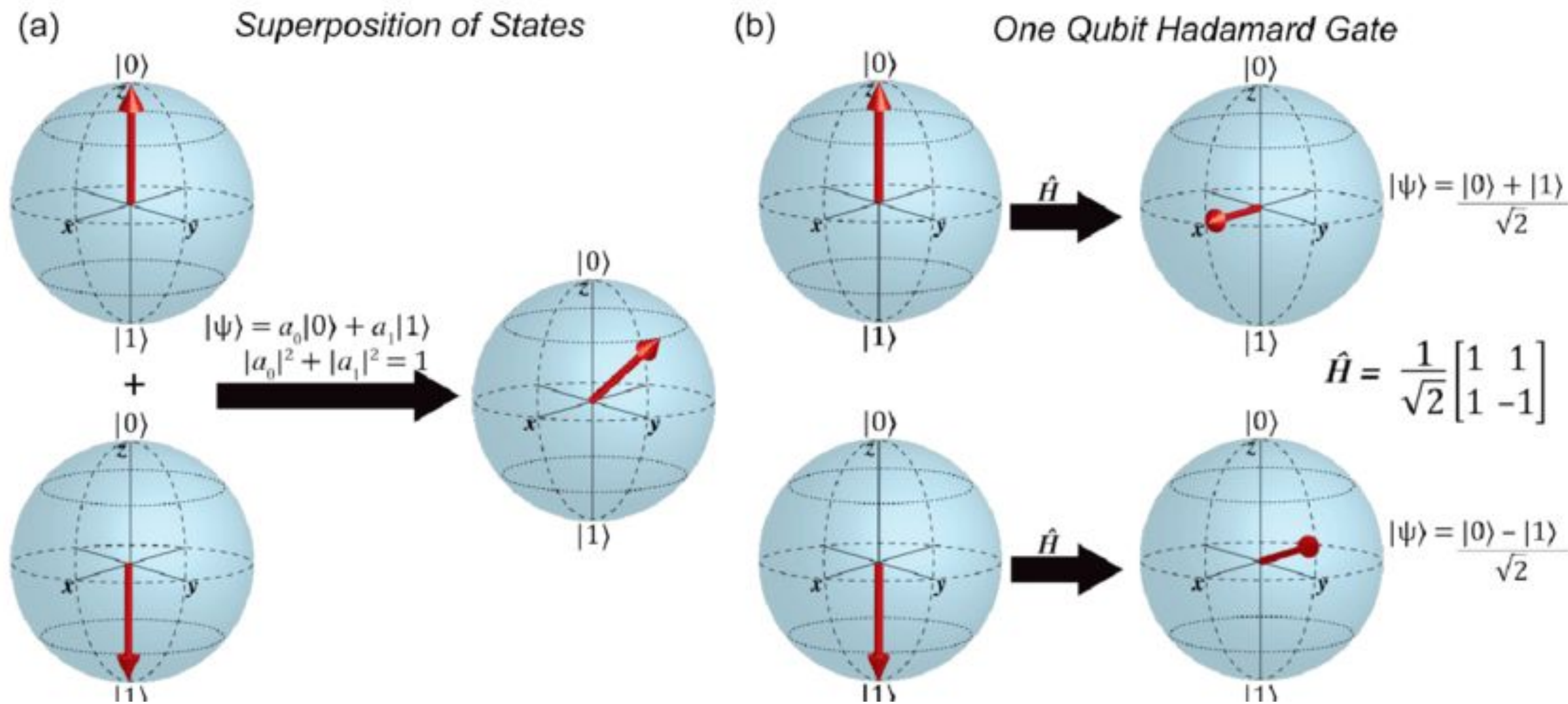
GANs

Generative Adversarial Nets



Quantum Machine Learning

Quantum Computing and Machine Learning

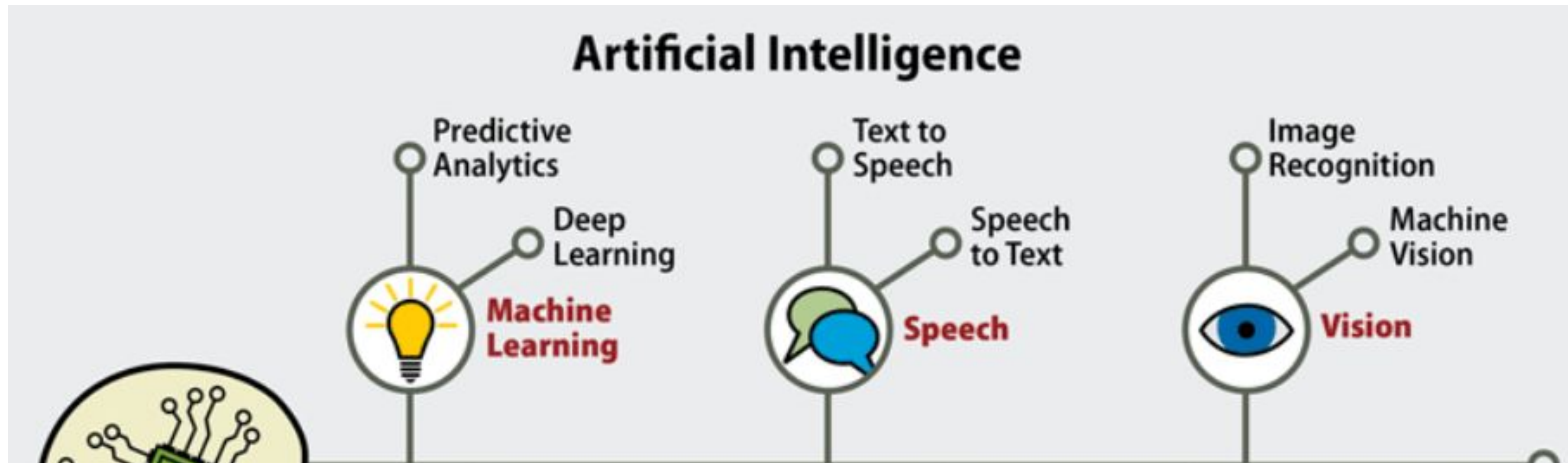


Ethics in A.I.



ML Career Planning

Choose a domain to focus





Elon Musk ✓
@elonmusk

...

Deus ex machine learning



6:07 AM · Nov 18, 2020 · Twitter for iPhone

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