



IntroToNeuralNetworks - IntroToTensorflow - 1

One should look for what is and not what he thinks should be. (Albert Einstein)

Intro to TensorFlow: Topic introduction

In this part of the course, we will cover the following concepts:

- Overview of TensorFlow / Keras building blocks
- Implement and fit a neural network model using Tensorflow on train data
- Evaluate neural network model on test data

Module completion checklist

Objective	Complete
Introduce TensorFlow and Keras	
Define a neural network model using Tensorflow	

Warm up: What is TensorFlow

- Have you heard of TensorFlow, an open source platform for machine learning?
- Watch the video **TensorFlow in 100 seconds** to get started and know more about TensorFlow
- After watching the video, share your thoughts on the following questions in the chat:
 - What feature was the most interesting?
 - What application was the most surprising?



TensorFlow

- **TensorFlow** is a base production-grade, **deep learning framework** for distributed processing and data flow graphs modeled on tensor data objects (originally from Google)
- Its core is in C++ with a layer of Python around it, abstracting processes into graphs and giving the core instructions
- *Click here* to get more information about TensorFlow



Keras

- Keras is a more high-level wrapper around TensorFlow, making it easier to build models that resemble other Python & scikit-learn functions you are used to
- It is a **deep learning library** explicitly built for Python
- It's a few layers up the abstraction chain from the core in C++, but it allows us to train models efficiently
- *Click [here](#)* to get more information about Keras
- **Note:** Since TensorFlow 2.0 fully integrated Keras, there is no need to install Keras separately



TensorFlow and Keras

- **TensorFlow** and **Keras** do **NOT** have to go together!
- Keras can be run on top of other systems like CNTK
- You can place your TensorFlow code directly into the Keras training pipeline or model
- If you want to develop your own algorithms, you can do so with TensorFlow



TensorFlow vs. Keras

- Although TensorFlow native modules alone can do everything (and then some) that Keras modules are built for, there are a few advantages of using Keras based on the four guiding principles used by Francois Chollet (the author of Keras):
 - **Modularity:** a model can be understood as a sequence or a graph alone; all the concerns of a deep learning model are discrete components that can be combined in arbitrary ways
 - **Minimalism:** the library provides just enough to achieve an outcome, no frills, and maximizing readability
 - **Extensibility:** new components are easy to add and use within the framework, intended for researchers to try and explore new ideas
 - **Python:** no separate model files with custom file formats as everything is native Python

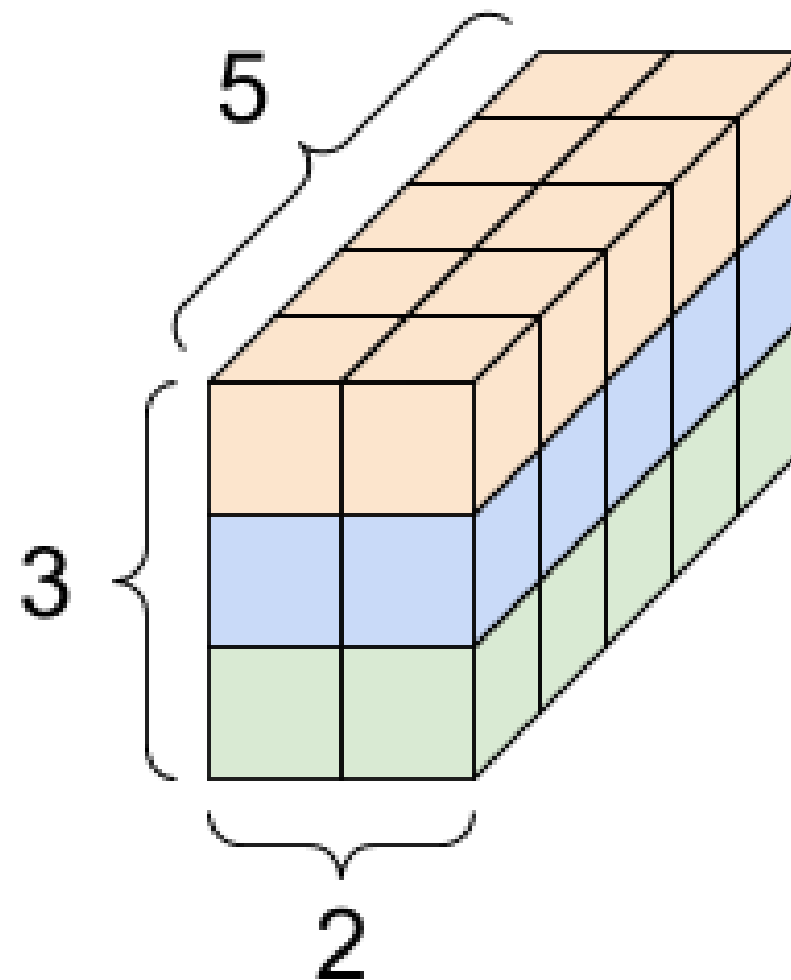
TensorFlow's APIs

- TensorFlow has APIs that support several languages including:
 - Python
 - JavaScript
 - C++
 - Java
- Additional community-supported languages include C#, Julia, Ruby, Rust, and Scala
- There are several available modules such as Lite and TensorFlow Extended (TFX)
- **Click [here](#)** to get more information about TensorFlow's APIs

Tensors

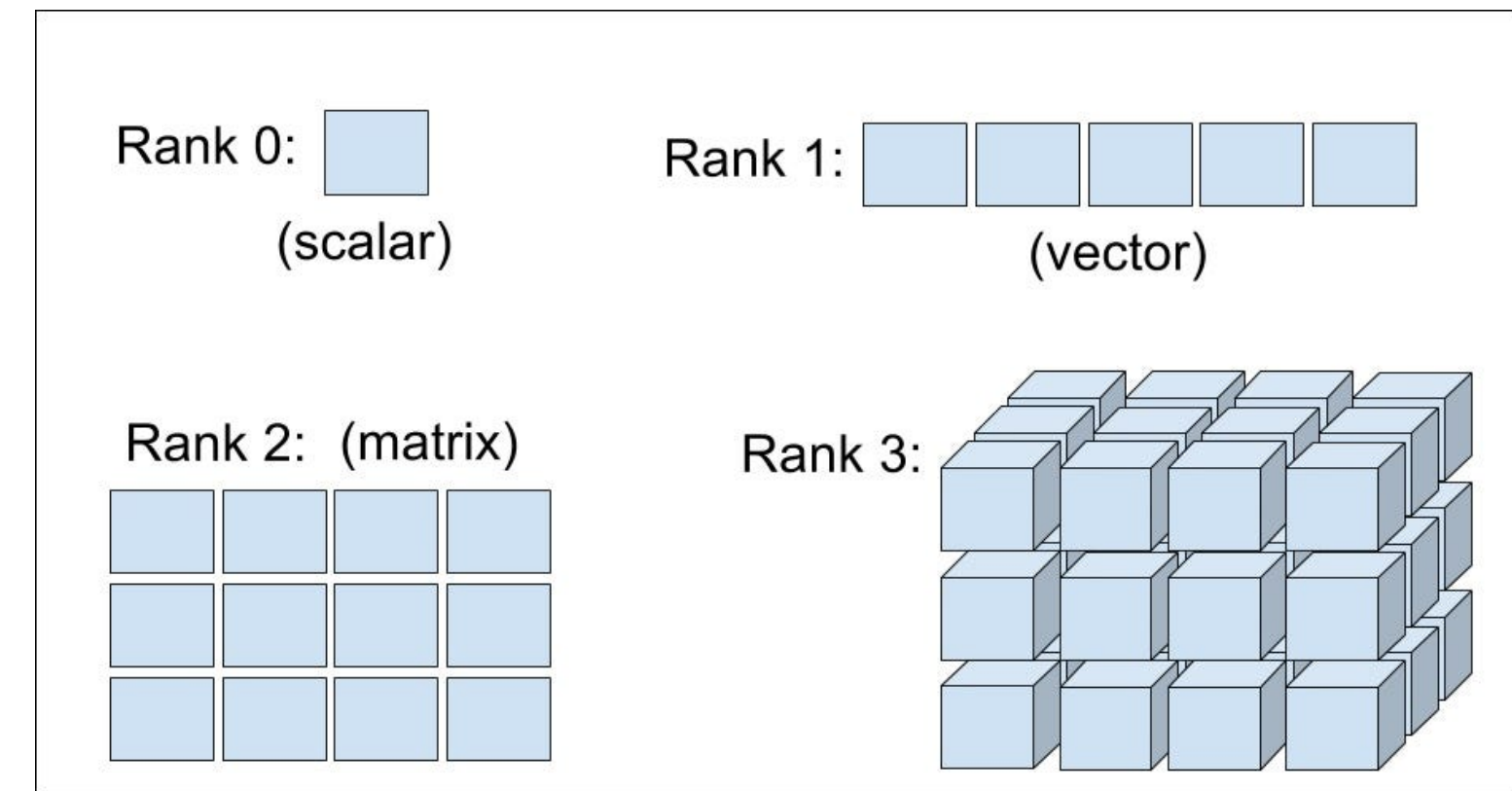
What are tensors?

- **Tensors** are multi-dimensional arrays with a uniform type (called a dtype)
- All tensors are immutable like Python numbers and strings: you can never update the contents of a tensor, only create a new one
- When working with TensorFlow, there is a high probability that you will end up working with tensors at some point



Tensor vocabulary

- Let's discuss important concepts associated to tensors:
 - **Shape:** The length (number of elements) of each of the dimensions of a tensor
 - **Rank:** Number of tensor dimensions
 - **Axis or Dimension:** A particular dimension of a tensor
 - **Size:** The total number of items in the tensor, the product shape vector



Tensor shape

- You may be familiar with scalars, vectors, matrices, and tensors, but let's discuss how these translate into TensorFlow

Tensor shape	Math type
0	scalar (magnitude)
1	vector (magnitude and direction)
2	matrix
3	3D-tensor (cube)
4+	multi-dimensional tensor (hypercube)

- [Click here](#) to get more information on creating basic tensors

Load TensorFlow for Python

- Loading TF is very simple, just run the following code:

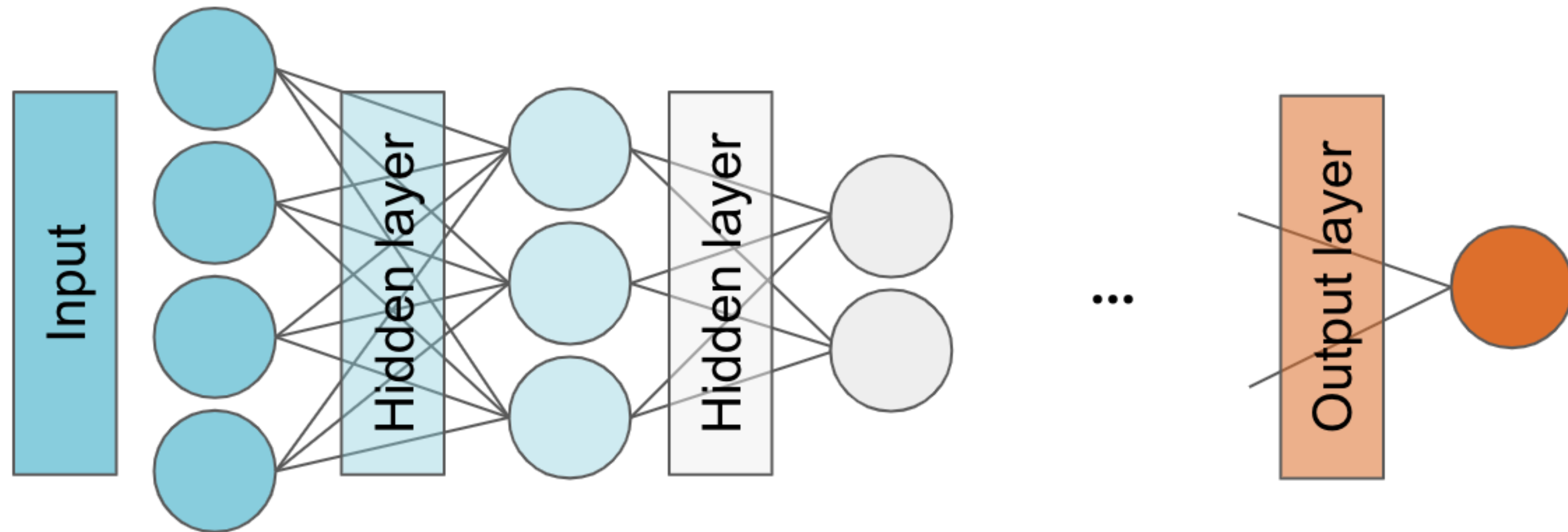
```
# Import tensorflow.  
import tensorflow as tf
```

- The `tf` library is the main TensorFlow package that is loaded into the environment
- **Click here** to get the list of all modules available through `tf`

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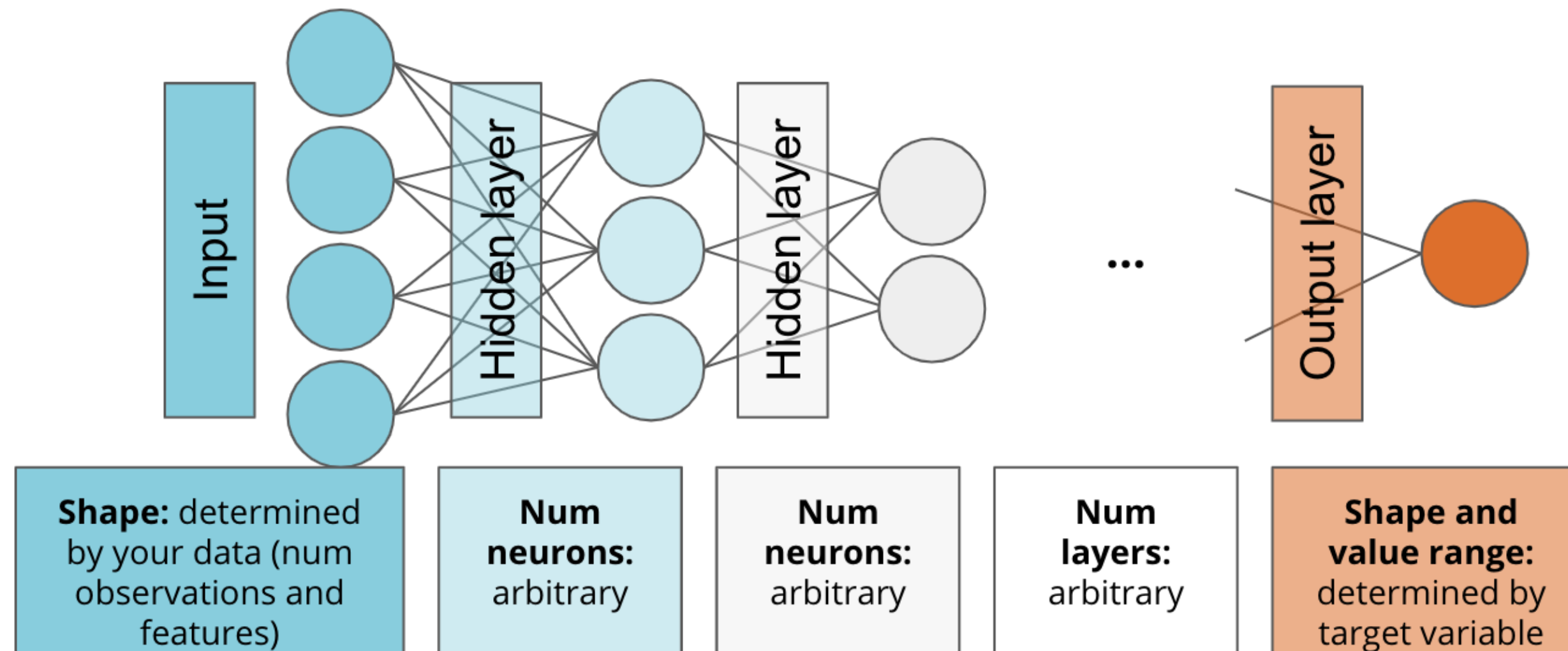
Neural network model: layers



- Any neural network modeling starts with the architecture
- We need 1 **input** and 1 **output** layer with at least 1 **hidden** layer in between
- The number of **hidden layers** determines the **depth of the network**
- It's a trial-and-error process; you must adjust it for every model you build

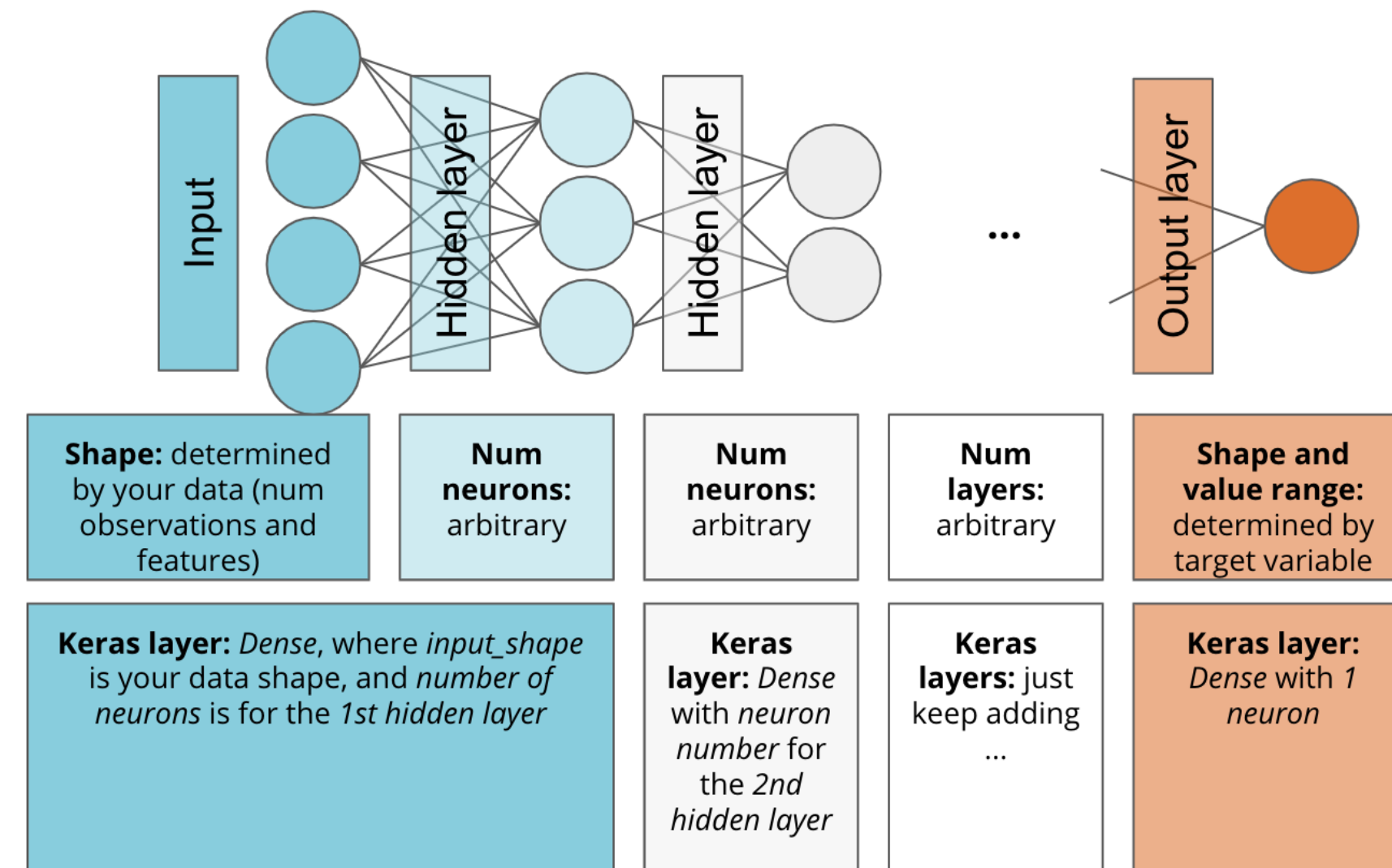
Neural network model: layers and neurons

- The **output** for a binary classification problem will have a **single neuron with a sigmoid** activation function

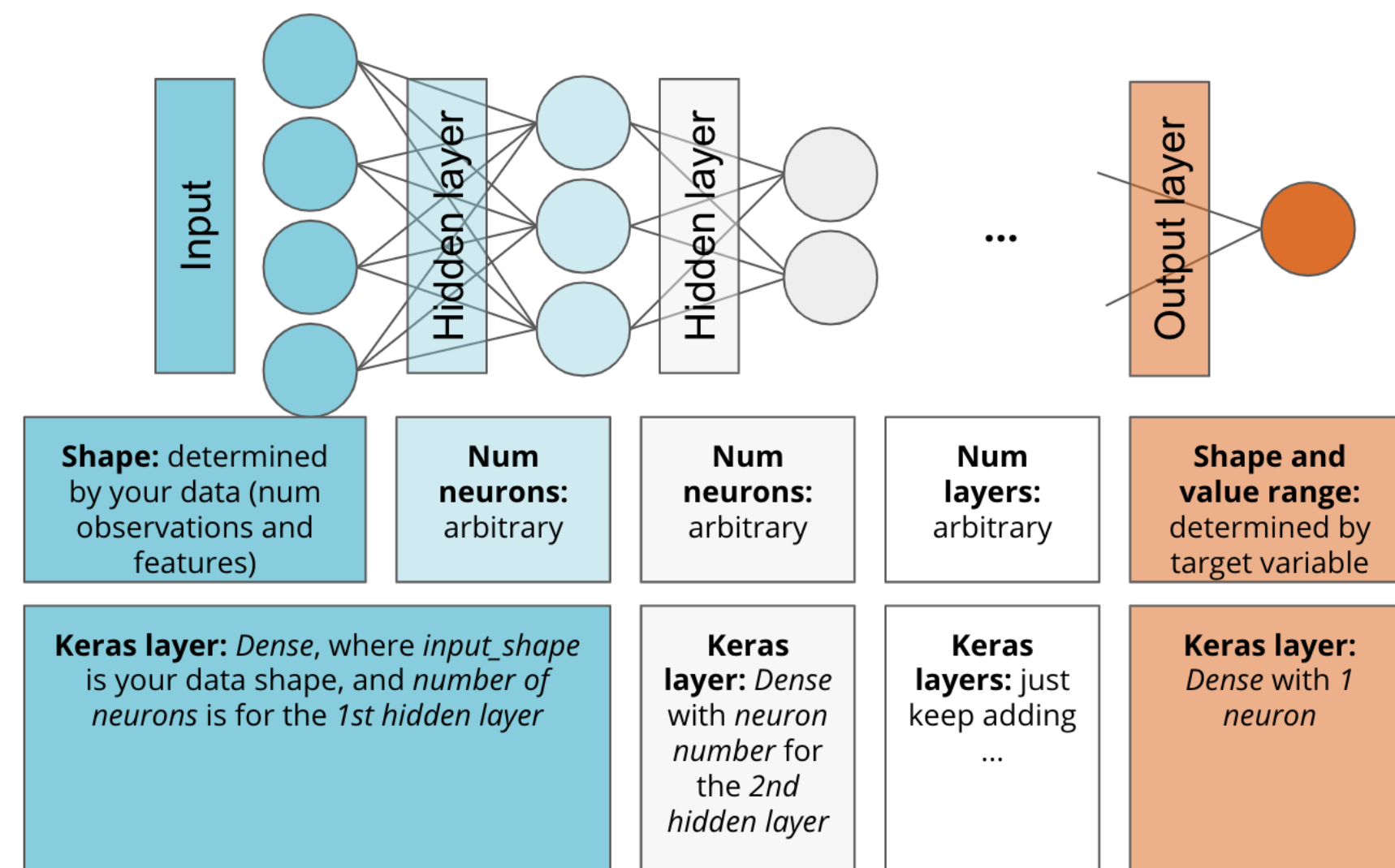


Neural network model: implemented in tf.keras

- Keras model architecture follows this logic:
 - choose model type
 - add your layers one-by-one and make sure to pick your parameters properly

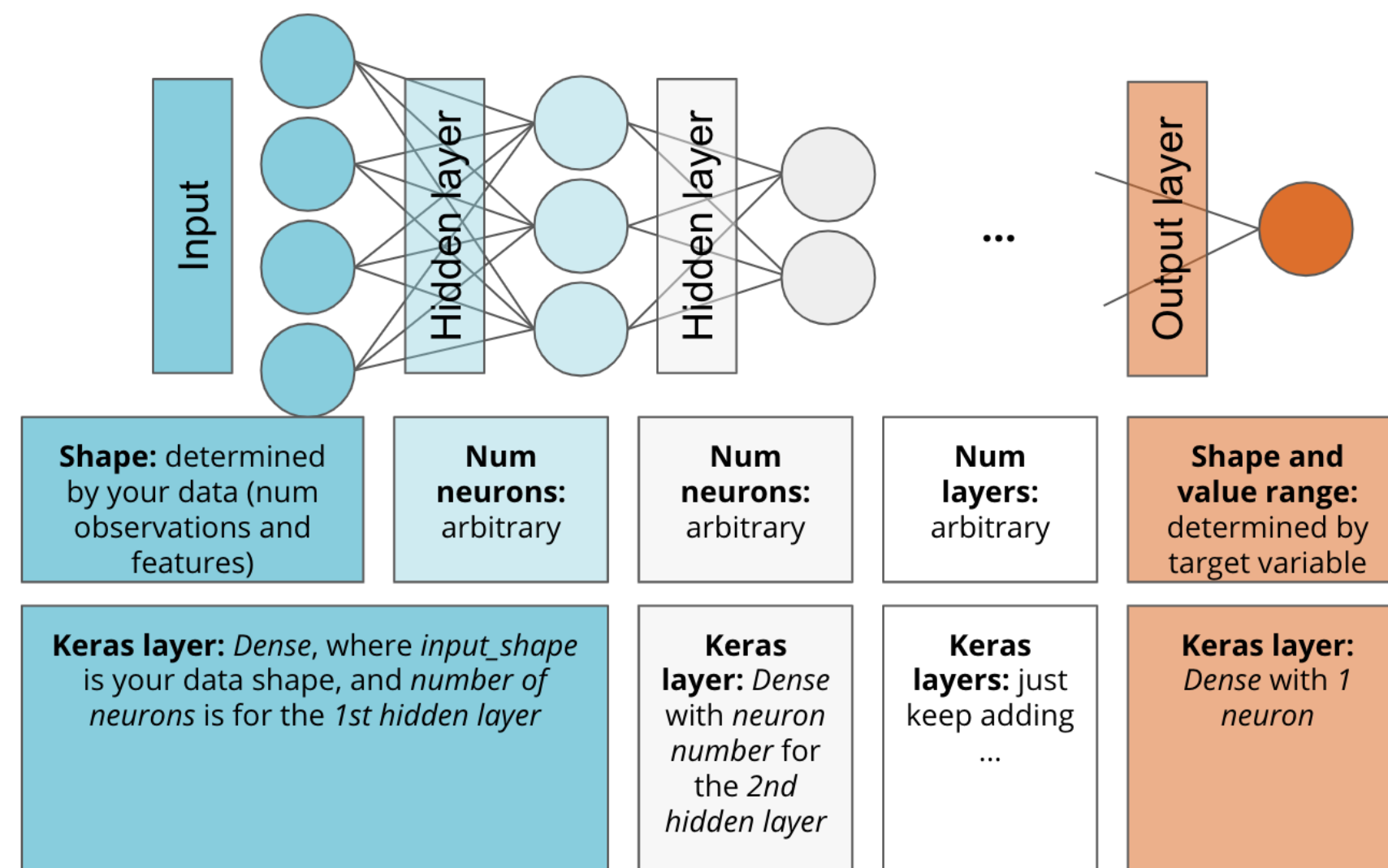


Neural network model: implemented in tf.keras (cont'd)



- There are two main types of models available in Keras:
 - **Sequential model:** A linear stack of layers with each layer having exactly one input and one output tensor
 - **Functional API:** Uses the same layers as the sequential model but provides more flexibility and can handle models with non-linear topology, shared layers, and multiple inputs or outputs

Neural network model: implemented in tf.keras (cont'd)



- We will discuss the `Sequential` model first, as we will start by working with one input tensor and one output tensor at each layer
 - Each layer will be `Dense`, i.e., densely connected to the preceding and following layers
 - For each layer we just need to pick appropriate parameters

Keras Sequential model with dense layers

- To create a `Sequential` model, you must pass a **list of layers** to the model constructor

```
Sequential([
    Dense()
    Dense()
    ...
])
```

- **Click here** for the package documentation on the TensorFlow website

TensorFlow > API > TensorFlow Core r2.1 > Python

tf.keras.Sequential

See Stable See Nightly

TensorFlow 1 version View source on GitHub

Class Sequential

Linear stack of layers.

Inherits From: [Model](#)

+ View aliases

Used in the notebooks

Used in the guide	Used in the tutorials
<ul style="list-style-type: none">Keras overviewMigrate your TensorFlow 1 code to TensorFlow 2Recurrent Neural Networks (RNN) with KerasDistributed training with TensorFlowEager execution	<ul style="list-style-type: none">Overfit and underfitTime series forecastingConvolutional Variational AutoencoderDeep Convolutional Generative Adversarial NetworkPix2Pix

Arguments:

- `layers`: list of layers to add to the model.

Keras Sequential model with dense layers (cont'd)



- Each `Dense` layer in the list will have the following form
- [Click here](#) for the package documentation on the TensorFlow website

```
Dense(units,                #<- num neurons  
      activation = None,    #<- activation  
      function  
      ...  
      )
```

TensorFlow > API > TensorFlow Core r2.1 > Python

tf.keras.layers.Dense

✓ See Stable See Nightly

 TensorFlow 1 version  View source on GitHub

Class Dense

Just your regular densely-connected NN layer.

Inherits From: [Layer](#)

Knowledge check



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Introduce TensorFlow and Keras	✓
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Congratulations on completing this module!

