

# Keras Functional Model-Non-linear Neural Networks

- The **Functional API** in Keras is a more flexible way to build neural networks than the simpler **Sequential** model.

```
from keras.models import Model
```

- 🙌 It needs **input** & **output** layers
  - There can be multiple input & output layers



## Difference from **Sequential** :

Feature	Sequential API	Functional API
Structure	Straight line	Any shape (branches, merges, etc.)
Suitable for	Simple models	Complex, custom models
Limitations	Only one input/output, no branching	Multiple inputs/outputs, shared layers, skip connections

## What is a Non-linear Neural Network?

A **non-linear neural network** is just a regular neural network that uses **non-linear activation functions** like:

- **ReLU**
- **tanh**
- **sigmoid**
- **softmax**

Without these, a neural network is just a **linear function**, which can't learn complex patterns.

## Why Use Functional API?

You **must** use Functional API if you want:

- **Multiple inputs or outputs**
- **Layers that share weights**
- **Skip connections (like in ResNet)**
- **Combining outputs from different layers**
- **Non-linear topologies (not just a straight line)**

## Python code



**You have to name each layer & state where it's getting input from.**

### Simple Structure:

**Define our input & output layers:**

```
from keras.models import Model

model = Model(inputs = x ,outputs = [output1,output2])
```

**Import all layers:**

```
from keras.layers import *
```

## Define input layer:

```
x = Input(shape=(3,))
```

- We have 3 columns in our input layer

## Make Dense & output layers:

```
hidden1 = Dense(128,activation='relu')(x)
```

```
hidden2 = Dense(64,activation='relu')(hidden1)
```

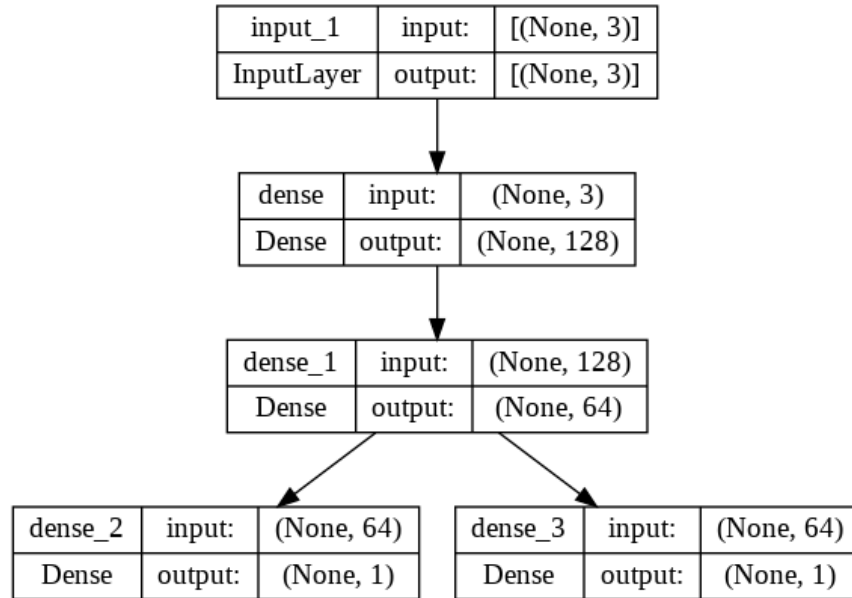
```
output1 = Dense(1,activation='linear')(hidden2)
```

```
output2 = Dense(1,activation='sigmoid')(hidden2)
```

- `(x)` , `(hidden1)` , etc. are **input tensors** to the current layer (output of the previous layer)
  - Meaning: `hidden1` is getting input from `x`
  - `hidden1` is getting input from `hidden2`

## Visualize this:

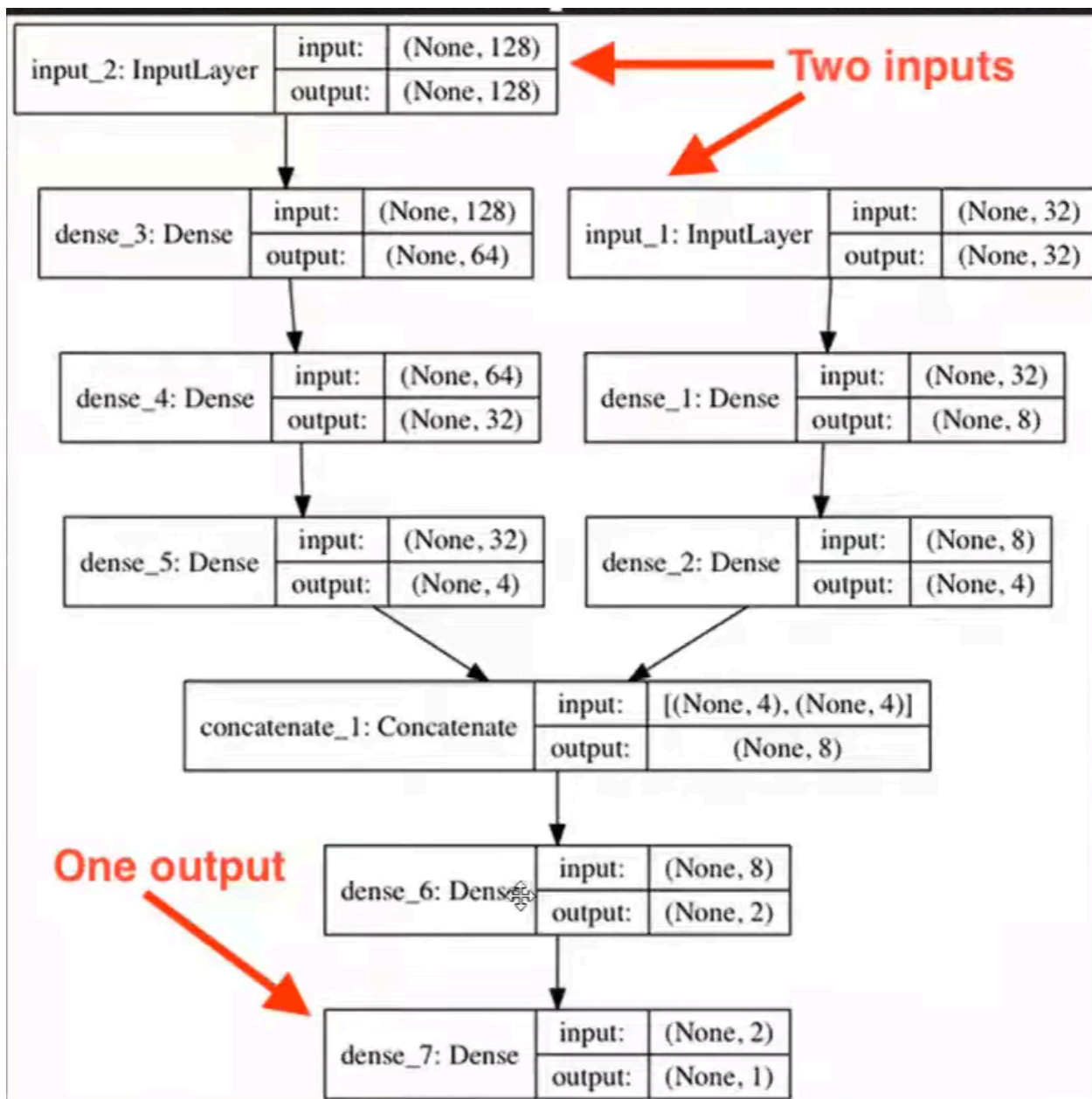
```
from keras.utils import plot_model  
plot_model(model,show_shapes=True)
```



## 2 Inputs, 1 Output:

### Flow:

1. Define **InputA** & **InputB**
2. There are 2 branches
  - a. Branch 1: Gets **InputA**
  - b. Branch 2: Gets **InputB**
3. **Combine (concatenate)** the output of the two branches
4. Make Dense & Output layer
5. Make a model



## 1. Define Input Layers:

```
inputA = Input(shape=(32,)) # First input (e.g., tabular data)
inputB = Input(shape=(128,)) # Second input (e.g., text embeddings)
```

- **Purpose:** Creates two separate input tensors for different data types.
- **Note:**

- `inputA` expects 32-dimensional vectors.
  - `inputB` expects 128-dimensional vectors.
- 

## 2. Branch 1: `inputA`

```
x = Dense(8, activation="relu")(inputA) # First dense layer
x1 = Dense(4, activation="relu")(x)    # Second dense layer
```

- **Flow:** `inputA` → Dense (8 units) → Dense (4 units).
  - **Output Shape:** `(None, 4)` (4-dimensional features).
- 

## 3. Branch 2: `inputB`

```
y = Dense(64, activation="relu")(inputB) # First dense layer
y1 = Dense(32, activation="relu")(y)    # Second dense layer
y2 = Dense(4, activation="relu")(y1)    # Final layer (matches Branch 1's output)
```

- **Flow:** `inputB` → Dense (64 units) → Dense (32 units) → Dense (4 units).
  - **Output Shape:** `(None, 4)` (4-dimensional features, same as Branch 1).
- 

## 4. Combine Branches (Concatenation)

```
combined = concatenate([x1, y2]) # Merges outputs from both branches
```

- **Purpose:** Combines features from both inputs into a single tensor.
  - **Output Shape:** `(None, 8)` (4 + 4 = 8 dimensions).
- 

## 5. Final Dense Layers for Prediction

```
z = Dense(2, activation="relu")(combined) # Intermediate layer
z1 = Dense(1, activation="linear")(z)    # Output layer (regression)
```

- **Flow:** Combined features → Dense (2 units) → Dense (1 unit).
- **Activation:**
  - `relu` for intermediate layer (non-linearity).
  - `linear` for output (regression task).

## 6. Define the Model

```
model = Model(inputs=[inputA, inputB], outputs=z1)
```

- **Inputs:** Accepts both `inputA` and `inputB` as a list.
- **Output:** Single regression value ( `z1` ).

## 7. Visualize the Model

```
from keras.utils import plot_model
plot_model(model, show_shapes=True)
```

- **Output:** A diagram showing the two input branches, concatenation, and output layers.

