

Copyright Notice

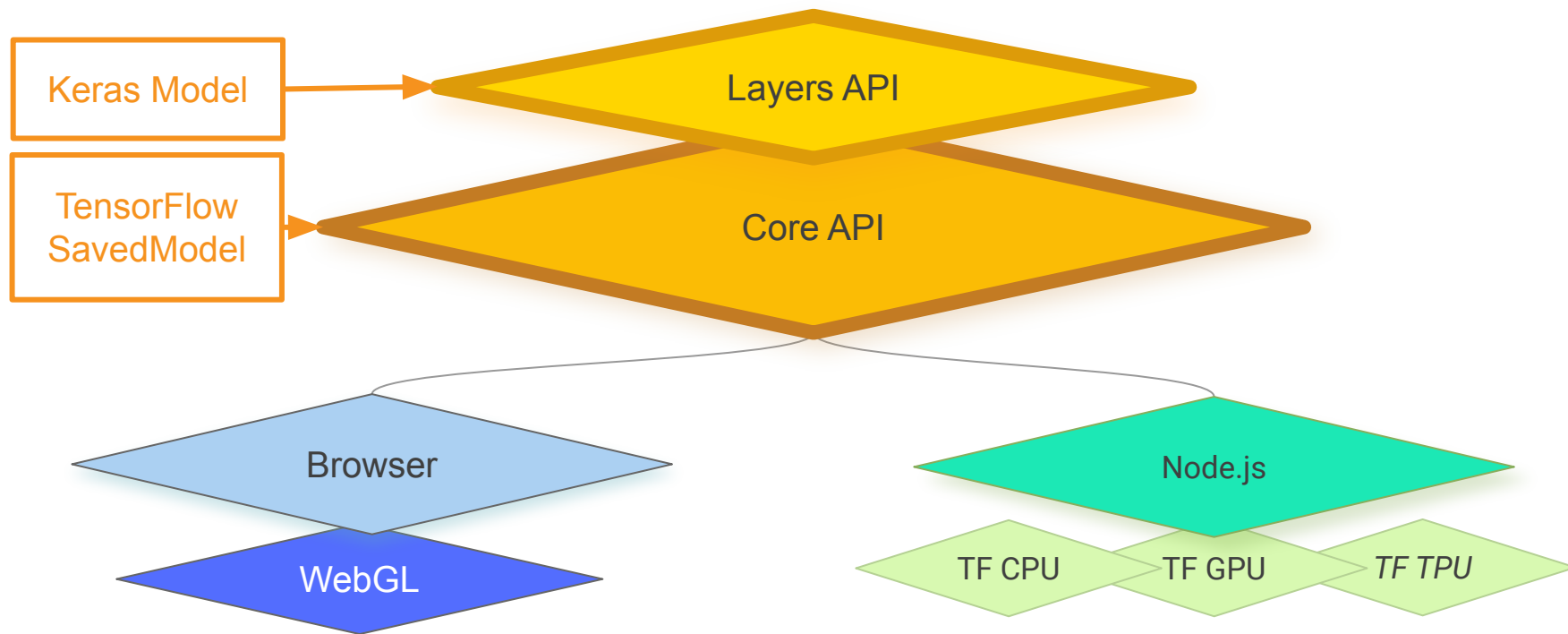
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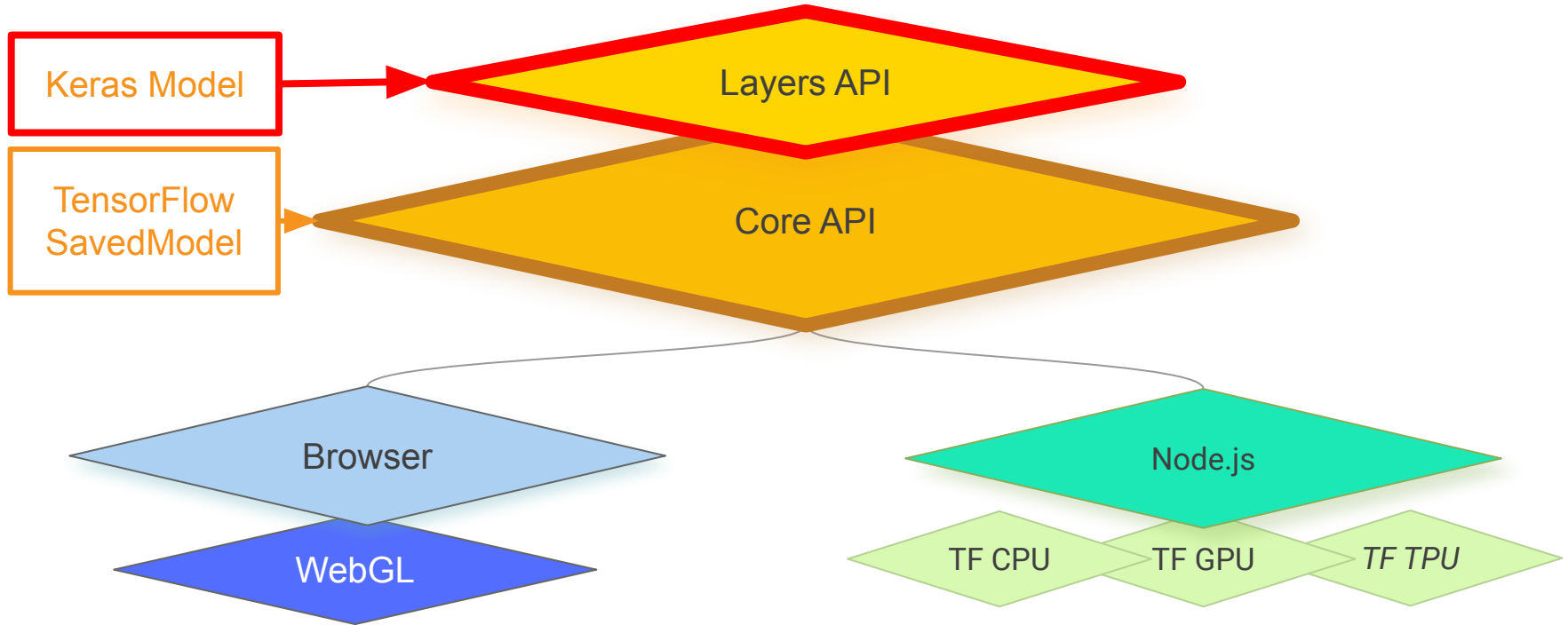
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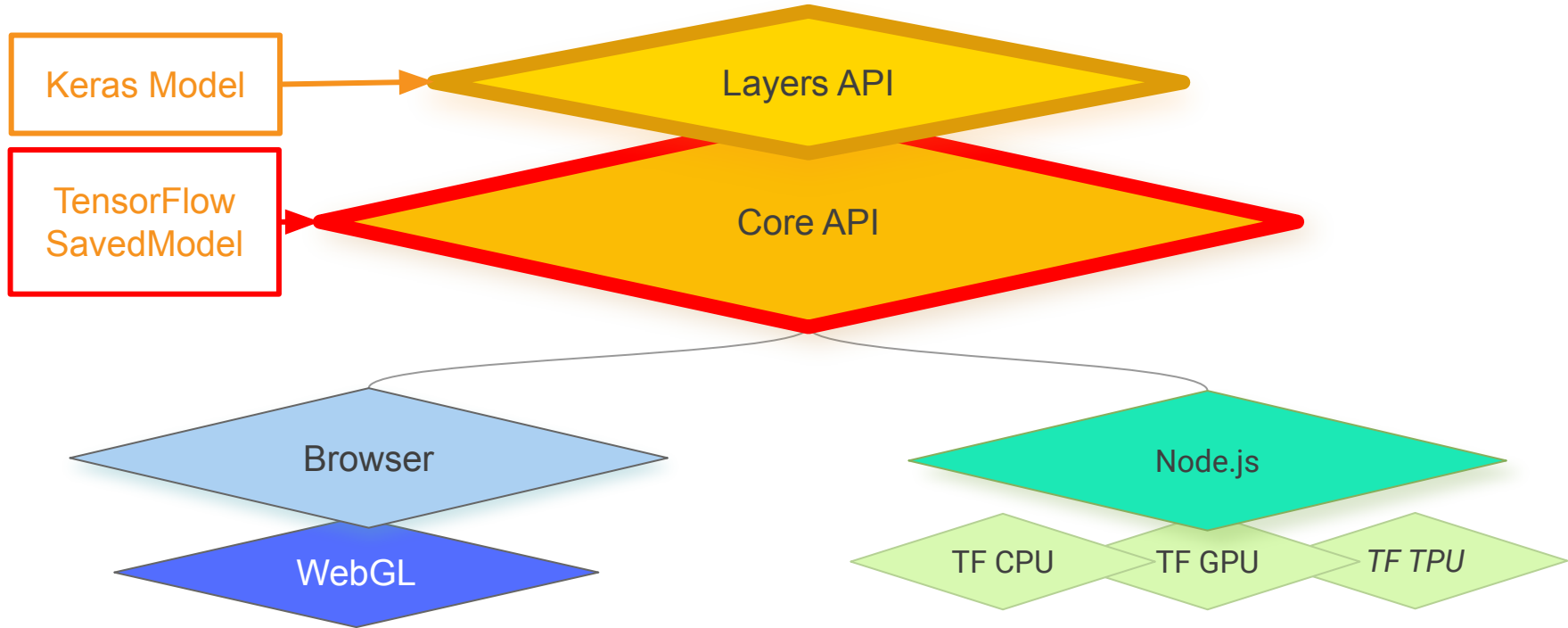
API



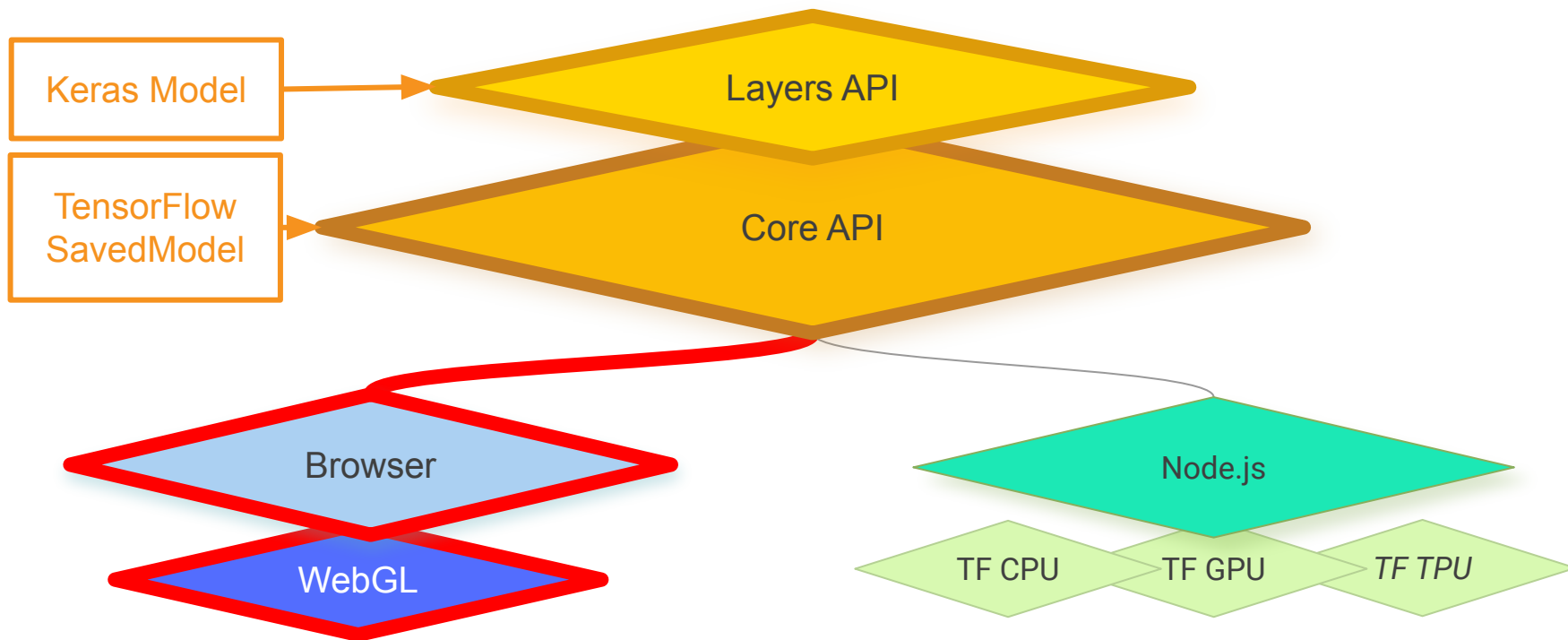
API



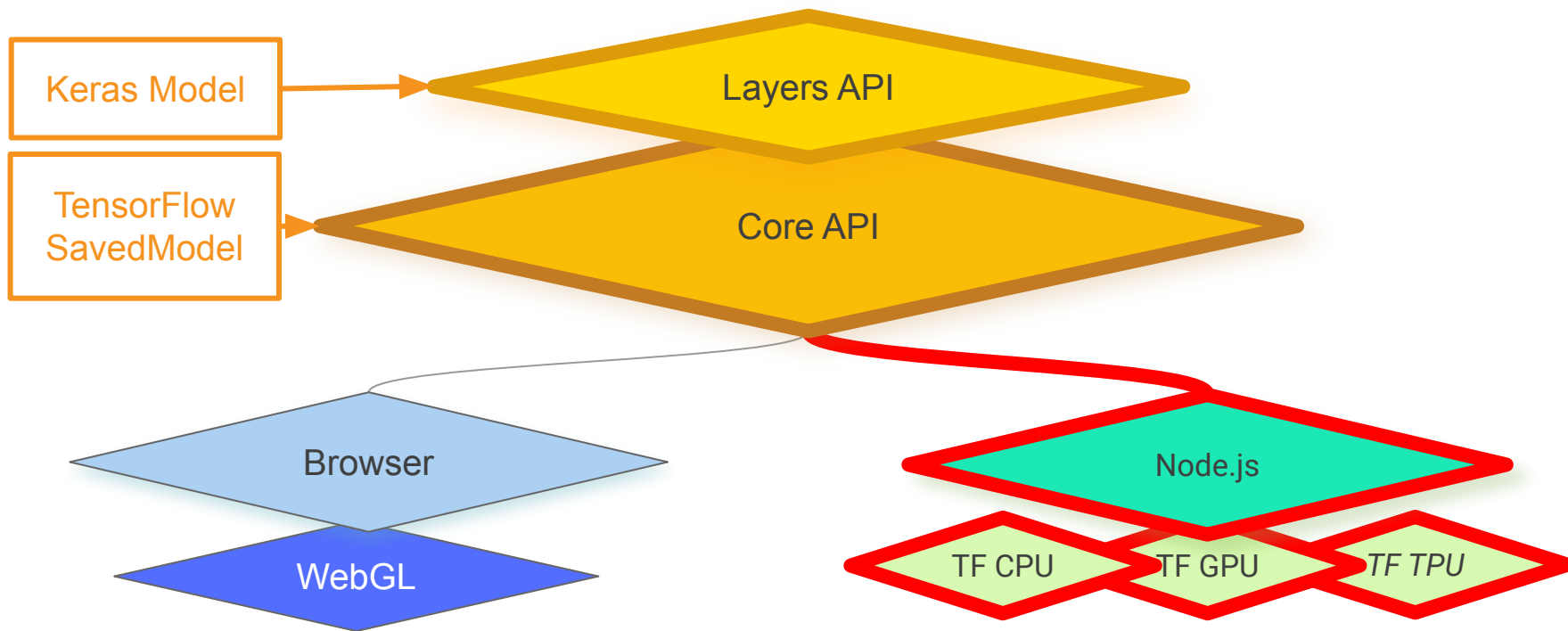
API



API



API



```
<html>  
<head></head>  
<body>  
    <h1>First HTML Page</h1>  
</body>  
</html>
```



```
<script lang="js">
  const model = tf.sequential();
  model.add(tf.layers.dense({units: 1, inputShape: [1]}));
  model.compile({loss: 'meanSquaredError',
                 optimizer: 'sgd'});

  model.summary();
</script>
```

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const model = tf.sequential();
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model.add(tf.layers.dense({units: 1, inputShape: [1]}));
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```

Layer (type)	Output shape	Param #
=====		
dense_Dense1 (Dense)	[null,1]	2
=====		
Total params: 2		
Trainable params: 2		
Non-trainable params: 0		

```
const xs = tf.tensor2d([-1.0, 0.0, 1.0, 2.0, 3.0, 4.0], [6, 1]);  
const ys = tf.tensor2d([-3.0, -1.0, 2.0, 3.0, 5.0, 7.0], [6, 1]);
```

```
const xs = tf.tensor2d([-1.0, 0.0, 1.0, 2.0, 3.0, 4.0], [6, 1]);  
const ys = tf.tensor2d([-3.0, -1.0, 2.0, 3.0, 5.0, 7.0], [6, 1]);
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const ys = tf.tensor2d([-3.0, -1.0, 2.0, 3.0, 5.0, 7.0], [6, 1]);
```

```
doTraining(model).then(() => {  
    alert(model.predict(tf.tensor2d([10], [1, 1])));  
});
```

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doTraining(model).then(() => {  
    alert(model.predict(tf.tensor2d([10], [1,1])));  
});
```

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doTraining(model).then(() => {  
    alert(model.predict(tf.tensor2d([10], [1, 1])));  
});
```

```
async function doTraining(model){
  const history =
    await model.fit(xs, ys,
      { epochs: 500,
        callbacks:{
          onEpochEnd: async(epoch, logs) =>{
            console.log("Epoch:" + epoch + " Loss:" + logs.loss);
          }
        }
      });
}
```

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      });  
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```

UCI



Machine Learning Repository

Center for Machine Learning and Intelligent Systems

Iris Data Set

Download: [Data Folder](#), [Data Set Description](#)

Abstract: Famous database; from Fisher, 1936



Data Set Characteristics:	Multivariate	Number of Instances:	150	Area:	Life
Attribute Characteristics:	Real	Number of Attributes:	4	Date Donated	1988-07-01
Associated Tasks:	Classification	Missing Values?	No	Number of Web Hits:	2512455

Source:

Creator:

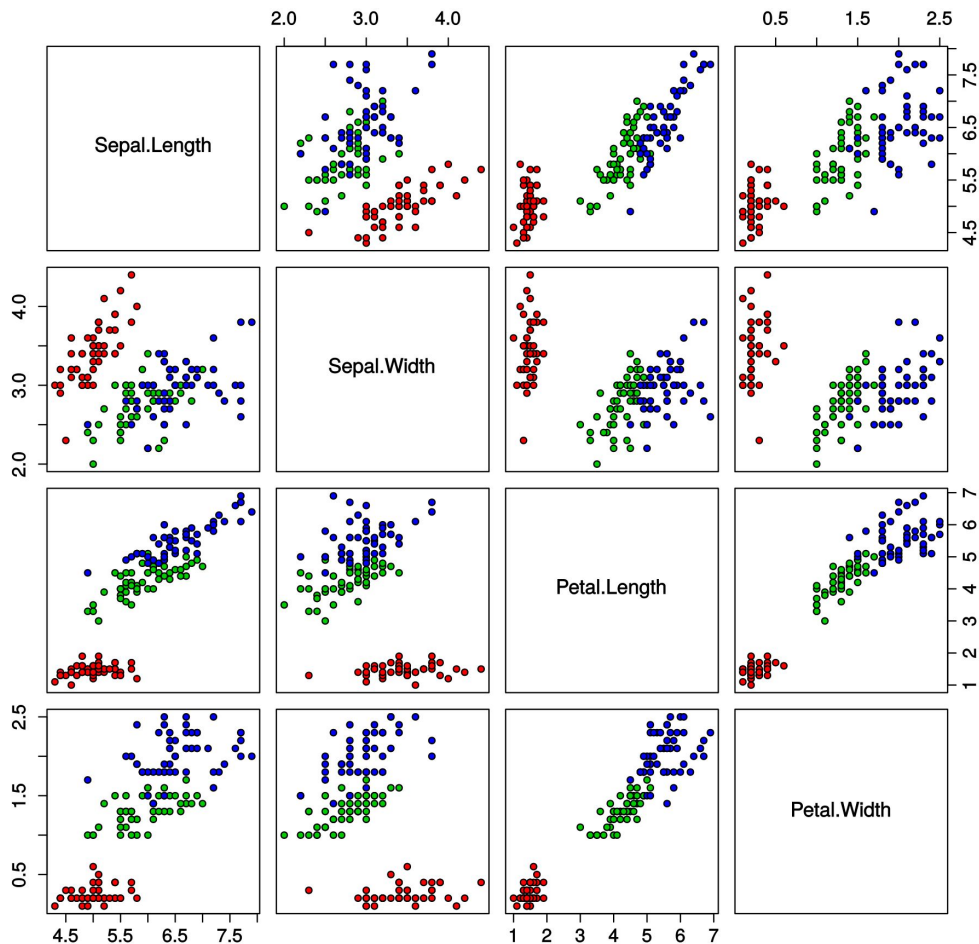
R.A. Fisher

Donor:

Michael Marshall (MARSHALL%PLU '@' io.arc.nasa.gov)

<https://archive.ics.uci.edu/ml/datasets/iris>

Iris Data (red=setosa,green=versicolor,blue=virginica)



By Nicoguardo - Own work, CC BY 4.0,
<https://commons.wikimedia.org/w/index.php?curid=46257808>

sepal_length, sepal_width, petal_length, petal_width, species

5.1, 3.5, 1.4, 0.2, setosa

4.9, 3, 1.4, 0.2, setosa

4.7, 3.2, 1.3, 0.2, setosa

4.6, 3.1, 1.5, 0.2, setosa

5, 3.6, 1.4, 0.2, setosa

5.4, 3.9, 1.7, 0.4, setosa

4.6, 3.4, 1.4, 0.3, setosa

5, 3.4, 1.5, 0.2, setosa


```
const csvUrl = 'iris.csv';  
const trainingData = tf.data.csv(csvUrl, {  
  columnConfigs: {  
    species: {  
      isLabel: true  
    }  
  }  
});
```

```
const csvUrl = 'iris.csv';
```

```
const trainingData = tf.data.csv(csvUrl, {  
  columnConfigs: {  
    species: {  
      isLabel: true  
    }  
  }  
});
```

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const csvUrl = 'iris.csv';  
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    }  
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  columnConfigs: {  
    species: {  
      isLabel: true  
    }  
  }  
});
```

```
const convertedData =  
  trainingData.map(({xs, ys}) => {  
    const labels = [  
      ys.species == "setosa" ? 1 : 0,  
      ys.species == "virginica" ? 1 : 0,  
      ys.species == "versicolor" ? 1 : 0 ]  
  
    return{ xs: Object.values(xs), ys:Object.values(labels)};  
  
  }).batch(10);
```

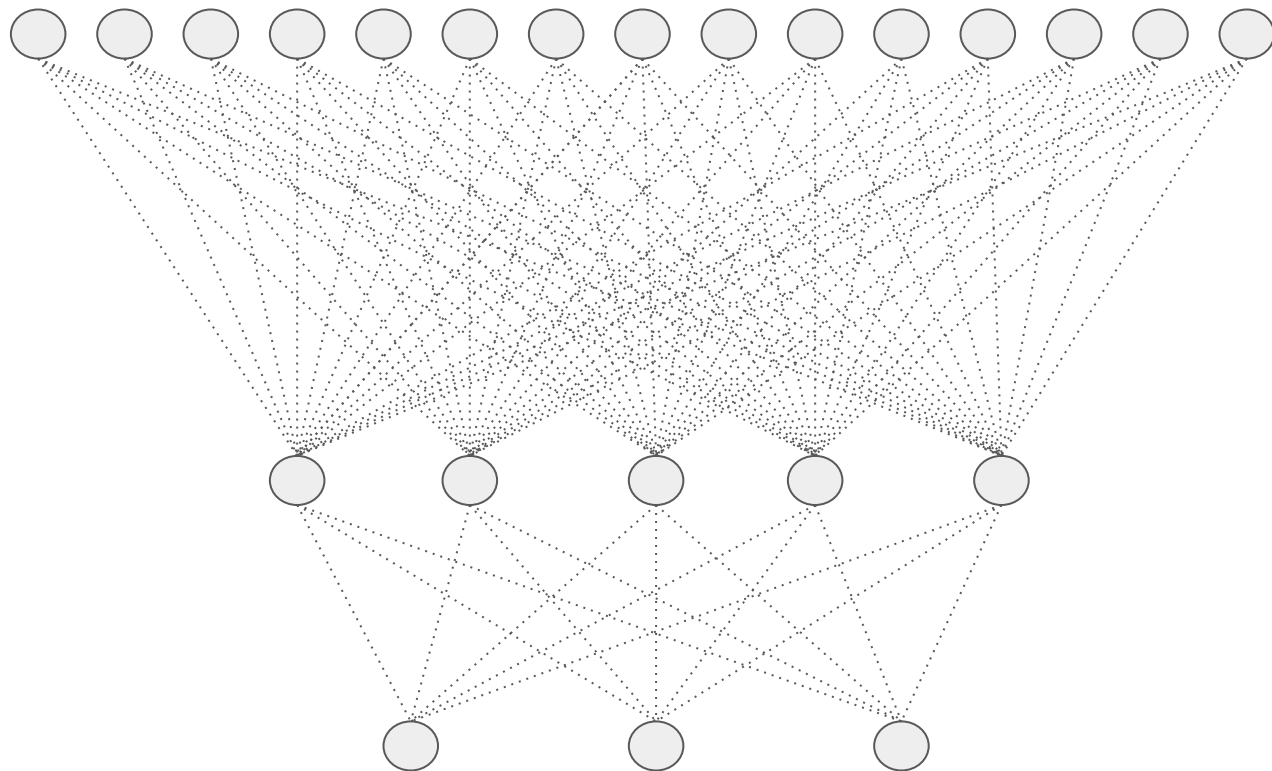
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const convertedData =  
trainingData.map(({xs, ys}) => {  
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```

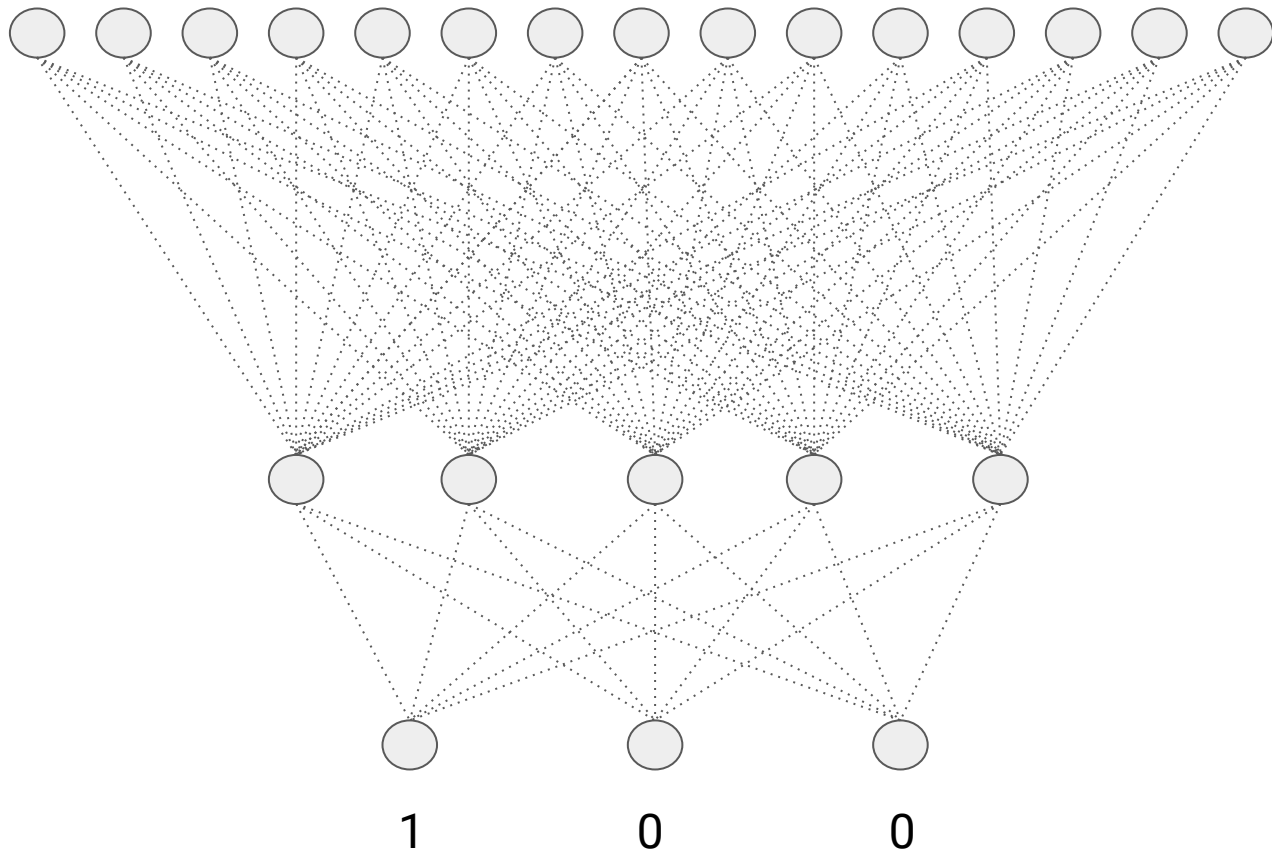
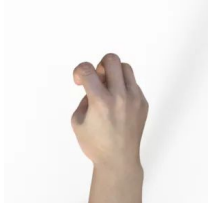
```
const convertedData =  
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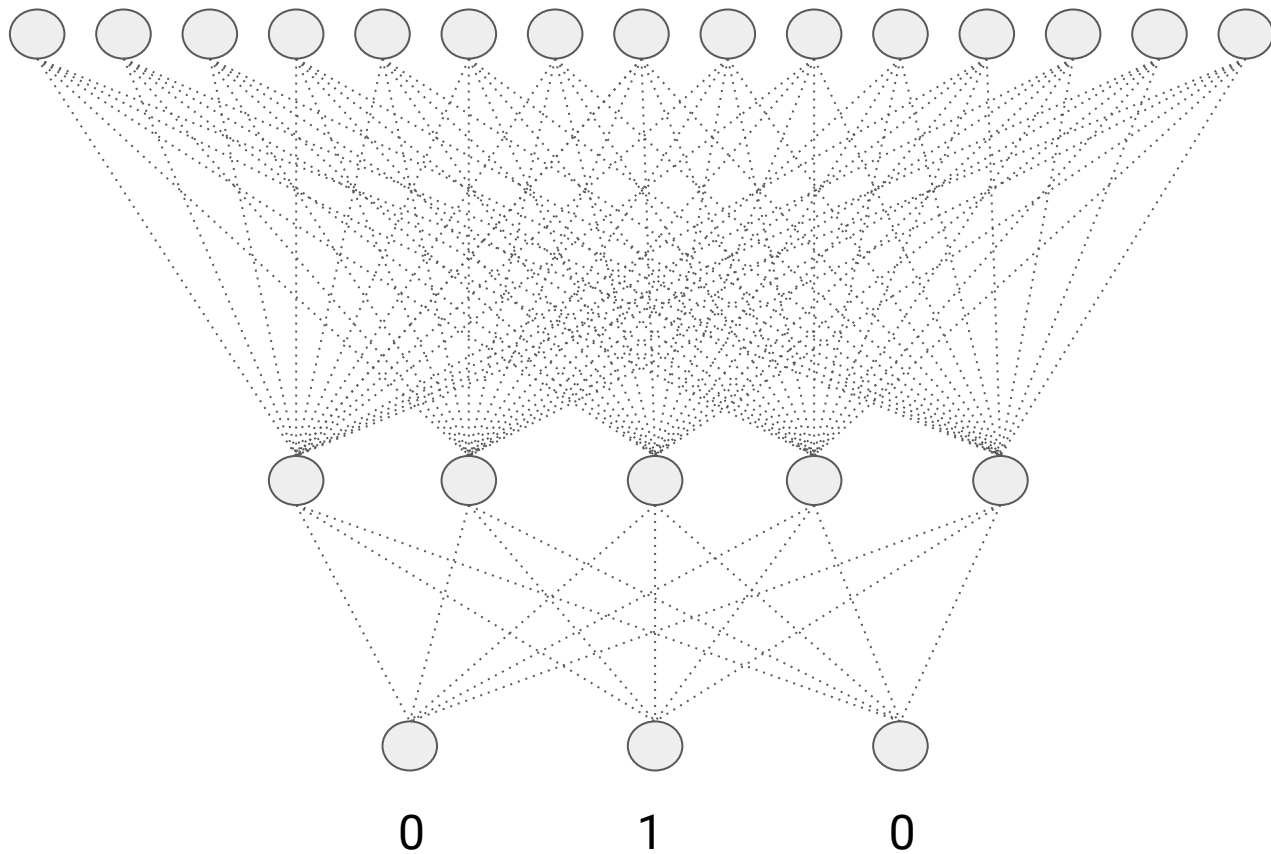
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const convertedData =  
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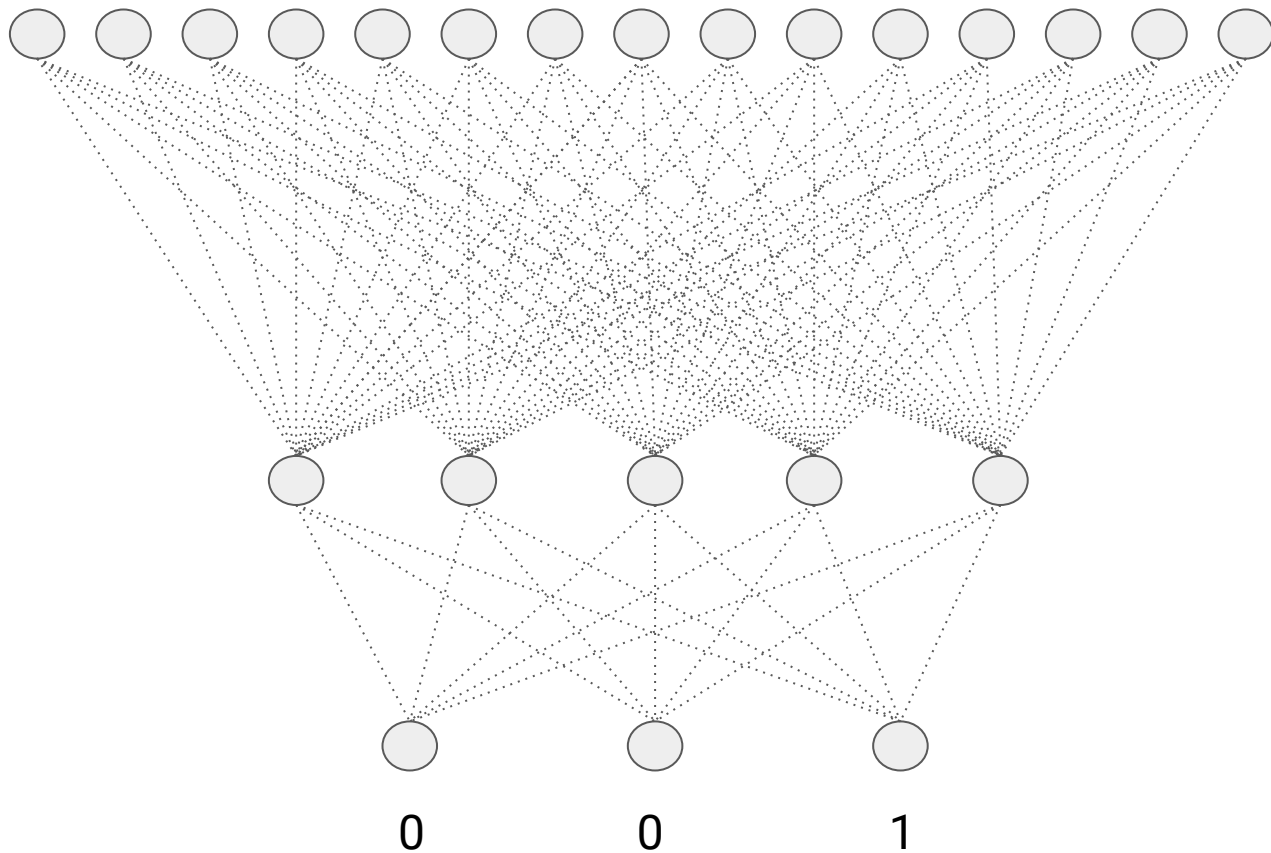


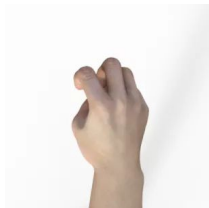
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const convertedData =  
  trainingData.map(({xs, ys}) => {  
    const labels = [  
      ys.species == "setosa" ? 1 : 0,  
      ys.species == "virginica" ? 1 : 0,  
      ys.species == "versicolor" ? 1 : 0 ]  
  
    return{ xs: Object.values(xs), ys:Object.values(labels)};  
  
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```











1

0

0



0

1




0



0

0

1

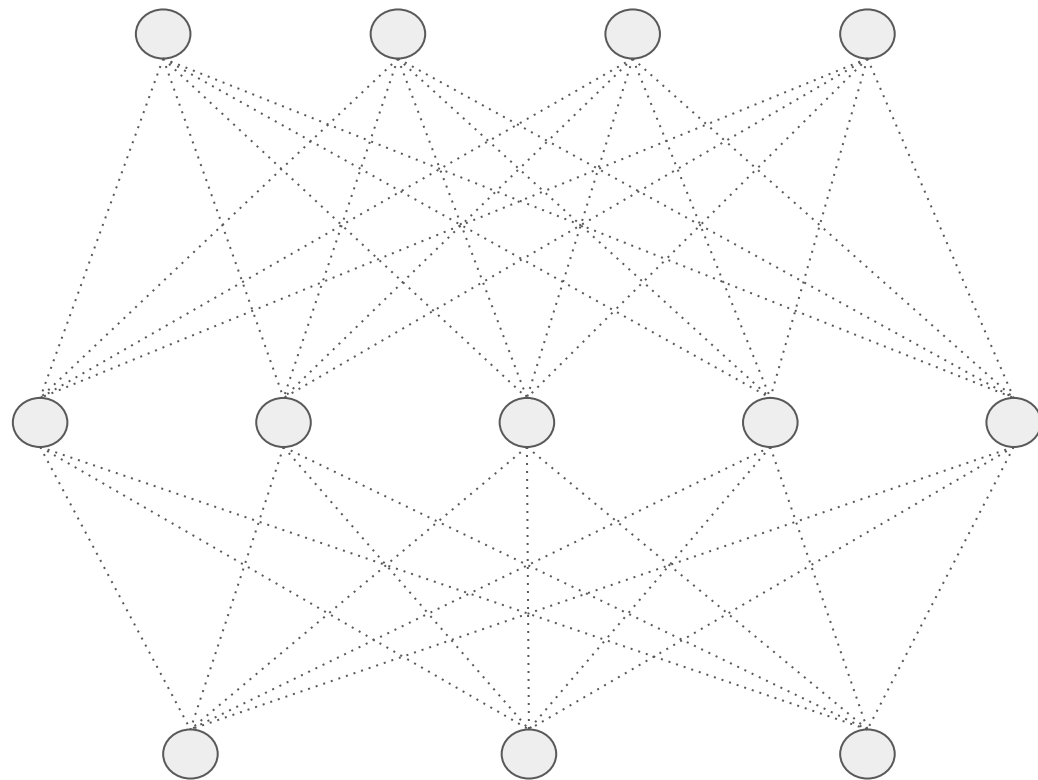
setosa		0	0
virginica	0		0
versicolor	0	0	

```
const convertedData =  
  trainingData.map(({xs, ys}) => {  
    const labels = [  
      ys.species == "setosa" ? 1 : 0,  
      ys.species == "virginica" ? 1 : 0,  
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    return{ xs: Object.values(xs), ys:Object.values(labels)};  
  
  }).batch(10);
```



setosa

virginica

versicolor

```
const model = tf.sequential();
```

```
model.add(tf.layers.dense({  
  inputShape: [numOfFeatures],  
  activation: "sigmoid", units: 5}))
```

```
model.add(tf.layers.dense({activation: "softmax", units: 3}));
```

```
const model = tf.sequential();
```

```
model.add(tf.layers.dense({  
  inputShape: [numOfFeatures],  
  activation: "sigmoid", units: 5}))
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```
model.add(tf.layers.dense({activation: "softmax", units: 3}));
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const model = tf.sequential();
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model.add(tf.layers.dense({  
  inputShape: [numOfFeatures],  
  activation: "sigmoid", units: 5}))
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```
model.add(tf.layers.dense({activation: "softmax", units: 3}));
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```
const model = tf.sequential();
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```
model.add(tf.layers.dense({  
  inputShape: [numOfFeatures],  
  activation: "sigmoid", units: 5}))
```

```
model.add(tf.layers.dense({activation: "softmax", units: 3}));
```

```
model.compile({  
    loss: "categoricalCrossentropy",  
    optimizer: tf.train.adam(0.06)}
```



```
await model.fitDataset(  
    convertedData,  
    {  
        epochs: 100,  
        callbacks: {  
            onEpochEnd: async(epoch, logs) => {  
                console.log("E: " + epoch + " Loss: " + logs.loss);  
            }  
        }  
    }  
));
```

```
await model.fitDataset(  
    convertedData,  
    {  
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        callbacks: {  
            onEpochEnd: async(epoch, logs) => {  
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            }  
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    }  
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            onEpochEnd: async(epoch, logs) =>{  
                console.log("E: " + epoch + " Loss: " + logs.loss);  
            }  
        }  
    });
```

```
const testVal = tf.tensor2d([5.8, 2.7, 5.1, 1.9], [1, 4]);  
const prediction = model.predict(testVal);  
alert(prediction);
```

```
const testVal = tf.tensor2d([5.8, 2.7, 5.1, 1.9], [1, 4]);  
const prediction = model.predict(testVal);  
alert(prediction);
```

```
const testVal = tf.tensor2d([5.8, 2.7, 5.1, 1.9], [1, 4]);  
const prediction = model.predict(testVal);  
alert(prediction);
```

```
const testVal = tf.tensor2d([5.8, 2.7, 5.1, 1.9], [1, 4]);  
const prediction = model.predict(testVal);  
alert(prediction);
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