# DEEP LEARNING GUIDE FOR DATA SCIENCE



Deep Learning is a subset of Machine Learning that uses neural networks to learn patterns from data. It's revolutionizing data science!

#### **NEURAL NETWORK ARCHITECTURE**

Neural networks consist of interconnected layers of nodes (neurons) that transform input data, passing it through each layer to produce an output. The network learns by adjusting weights during training.

```
# Sample code for a basic neural network
import tensorflow as tf

model = tf.keras.Sequential([
    tf.keras.layers.Dense(units=64, activation='relu', input_shape=
(input_dim,)),
    tf.keras.layers.Dense(units=32, activation='relu'),
    tf.keras.layers.Dense(units=output_dim, activation='softmax')
])
```



#### **ACTIVATION FUNCTIONS**

Activation functions introduce non-linearity, enabling neural networks to learn complex patterns and make predictions. ReLU is widely used for hidden layers due to its simplicity and effectiveness.

```
# Example code for ReLU activation function
def relu(x):
    return max(0, x)
```





#### **BACKPROPAGATION**

Backpropagation is the core algorithm behind training neural networks. It calculates gradients to adjust weights, minimizing errors and improving model accuracy during training.

# Formula for updating weights with backpropagation new\_weight = old\_weight - learning\_rate \* gradient





## CONVOLUTIONAL NEURAL NETWORKS (CNN)

CNNs are designed to process grid-like data, like images. They employ convolutional layers to detect features and pooling layers to reduce spatial dimensions, making them ideal for image recognition tasks.

```
# Code for a simple CNN
import tensorflow as tf

model = tf.keras.Sequential([
    tf.keras.layers.Conv2D(32, (3, 3), activation='relu',
input_shape=(width, height, channels)),
    tf.keras.layers.MaxPooling2D((2, 2)),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(64, activation='relu'),
    tf.keras.layers.Dense(num_classes, activation='softmax')
])
```





#### RECURRENT NEURAL NETWORKS (RNN)

RNNs excel at processing sequential data, like time series or natural language. Their recurrent connections allow them to maintain context and remember information from past inputs.

```
# Code for a basic RNN
import tensorflow as tf

model = tf.keras.Sequential([
    tf.keras.layers.SimpleRNN(64, activation='relu', input_shape=
(timesteps, features)),
    tf.keras.layers.Dense(num_classes, activation='softmax')
])
```





#### **REGULARIZATION TECHNIQUES**

Regularization methods like Dropout and L2 Regularization prevent neural networks from overfitting by reducing the impact of certain neurons during training and adding penalties to large weights.

```
# Example code for applying Dropout
import tensorflow as tf

model = tf.keras.Sequential([
    tf.keras.layers.Dense(64, activation='relu', input_shape=
(input_dim,)),
    tf.keras.layers.Dropout(0.25),
    tf.keras.layers.Dense(32, activation='relu'),
    tf.keras.layers.Dropout(0.25),
    tf.keras.layers.Dropout(0.25),
    tf.keras.layers.Dense(output_dim, activation='softmax')
])
```





#### <u>OPTIMIZERS</u>

Optimizers control how neural networks update their weights during training. Algorithms like Adam and RMSprop help speed up convergence and improve training efficiency.

```
# Example code using the Adam optimizer optimizer = tf.keras.optimizers.Adam(learning_rate=0.001) model.compile(optimizer=optimizer, loss='categorical_crossentropy', metrics=['accuracy'])
```





#### LOSS FUNCTIONS

Loss functions measure the difference between predicted and actual values during training. Choosing the right loss function is crucial, as it affects how the model learns from data.

# Example code for using categorical cross-entropy loss model.compile(optimizer='adam', loss='categorical\_crossentropy', metrics=['accuracy'])



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