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Department of Computer Science Engineering (AIML)

Deep Learning Lab

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AIM - To implement and evaluate the performance of the following neural network architectures with 2 hidden layers on the SONAR dataset for 200 epochs.

- 1. Backpropagation NN with adam optimizer.
- 2. I1 and I2 regularizations.
- 3. early stopping with p = 5
- 4. dropout regularization with p = 0.2 for input and second hidden layer.

Importing the Dependancies

```
import tensorflow as tf
from sklearn.datasets import make_classification
from sklearn.model_selection import train_test_split
```

Loading Our Data

```
# Load the Sonar dataset
X, y = make_classification(n_samples=208, n_features=60, n_informative=60, n_redundant=0,
# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

Creating a Sequential Model

```
%%time
nn = tf.keras.models.Sequential([
 tf.keras.layers.Dense(32, input_dim=60, activation='relu'),
 tf.keras.layers.Dense(16, activation='relu'),
 tf.keras.layers.Dense(8, activation='relu'),
 tf.keras.layers.Dense(1, activation='sigmoid')
])
nn.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
nn.fit(X_train, y_train, epochs=200)
Epoch 1/200
Epoch 2/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
Epoch 6/200
Epoch 7/200
```

6/6 [===============] - Os 2ms/step - loss: 0.3902 - accuracy: 0.8373

<keras.callbacks.History at 0x7f8263244970>

Epoch 8/200

Epoch 9/200

Epoch 10/200

```
%%time
nn_with_reg = tf.keras.models.Sequential([
  tf.keras.layers.Dense(32, input_dim=60, activation='relu', kernel_regularizer=tf.keras
  tf.keras.layers.Dense(16, activation='relu',kernel_regularizer=tf.keras.regularizers.l
  tf.keras.layers.Dense(8, activation='relu', kernel_regularizer=tf.keras.regularizers.)
  tf.keras.layers.Dense(1, activation='sigmoid', kernel_regularizer=tf.keras.regularizer
])
nn_with_reg.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
nn_with_reg.fit(X_train, y_train, epochs=200)
Epoch 1/200
Epoch 2/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
Epoch 6/200
Epoch 7/200
```

<keras.callbacks.History at 0x7f8262a8c9d0>

Epoch 8/200

Epoch 9/200

Epoch 10/200

```
%%time
nn_with_earlystopping = tf.keras.models.Sequential([
  tf.keras.layers.Dense(32, input_dim=60, activation='relu'),
  tf.keras.layers.Dense(16, activation='relu'),
  tf.keras.layers.Dense(8, activation='relu'),
  tf.keras.layers.Dense(1, activation='sigmoid')
])
early_stop = tf.keras.callbacks.EarlyStopping(monitor='val_loss', patience=5, restore_best
nn_with_earlystopping.compile(loss='binary_crossentropy', optimizer='adam', metrics=['acci
history = nn_with_earlystopping.fit(X_train, y_train, epochs=200, batch_size=8, validation
Epoch 1/200
Epoch 2/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
```

```
%%time
nn_with_dropout = tf.keras.models.Sequential([
    tf.keras.layers.Dense(32, input_dim=60, activation='relu'),
    tf.keras.layers.Dropout(0.2),
    tf.keras.layers.Dense(16, activation='relu'),
    tf.keras.layers.Dense(8, activation='relu'),
    tf.keras.layers.Dropout(0.2),
    tf.keras.layers.Dense(1, activation='sigmoid')
])
nn_with_dropout.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy']
nn_with_dropout.fit(X_train, y_train, epochs=200)
Epoch 1/200
```

```
Epoch 2/200
Epoch 3/200
Epoch 4/200
Epoch 5/200
Epoch 6/200
Epoch 7/200
Epoch 8/200
Epoch 9/200
```

<keras.callbacks.History at 0x7f825c7053d0>