





- a. Data preparation
- b. Generate training data
- c. Train model
- d. Output
- [1]: import numpy as np import re
- [2]: data = " ""Deep learning (also known as deep structured learning) is part of a broader family of machine learning methods based on artificial neural networ
- [2]: 'Deep learning (also known as deep structured learning) is part of a broader family of machine learning methods based on artificial neural networks with representation learning. Learning can be supervised, semi-supervised or unsupervised. Deep-learning architectures such as deep neural networks, deep beli ef networks, deep reinforcement learning, recurrent neural networks, convolutional neural networks and Transformers have been applied to fields including computer vision, speech recognition, natural language processing, machine translation, bioinformatics, drug design, medical image analysis, climate scien ce, material inspection and board game programs, where they have produced results comparable to and in some cases surpassing human expert performance.'
- [3]: sentences = data.split('.')
- [3]: ['Deep learning (also known as deep structured learning) is part of a broader family of machine learning methods based on artificial neural networks with epresentation learning', ' Learning can be supervised, semi-supervised or unsupervised',

  - Deep-learning architectures such as deep neural networks, deep belief networks, deep reinforcement learning, recurrent neural networks, convolutional neural networks and Transformers have been applied to fields including computer vision, speech recognition, natural language processing, machine translation, bioinformatics, drug design, medical image analysis, climate science, material inspection and board game programs, where they have produced results comparable to and in some cases surpassing human expert performance',

```
[4]: clean sent=[]
        for sentence in sentences:
              if sentence==""
                   continue
              sentence = re.sub('[^A-Za-z0-9]+', ' ', (sentence))
sentence = re.sub(r'(?:^| )\w (?:$| )', ' ', (sentence))
sentence = sentence.lower()
                                                                             ', (sentence)).strip()
              clean_sent.append(sentence)
```

[4]: ['deep learning also known as deep structured learning is part of a broader family of machine learning methods based on artificial neural networks with r epresentation learning',
'learning can be supervised semi supervised or unsupervised',

deep learning architectures such as deep neural networks deep belief networks deep reinforcement learning recurrent neural networks convolutional neural networks and transformers have been applied to fields including computer vision speech recognition natural language processing machine translation bioi nformatics drug design medical image analysis climate science material inspection and board game programs where they have produced results comparable to and in some cases surpassing human expert performance']

[5]: from tensorflow.keras.preprocessing.text import Tokenizer

```
[6]: tokenizer = Tokenizer()
     tokenizer.fit_on_texts(clean_sent)
     sequences = tokenizer.texts_to_sequences(clean_sent)
     print(sequences)
```

[[2, 1, 12, 13, 6, 2, 14, 1, 15, 16, 7, 17, 18, 19, 7, 8, 1, 20, 21, 22, 23, 4, 3, 24, 25, 1], [1, 26, 27, 9, 28, 9, 29, 30], [2, 1, 31, 32, 6, 2, 4, 3, 2, 33, 3, 2, 34, 1, 35, 4, 3, 36, 4, 3, 5, 37, 10, 38, 39, 11, 40, 41, 42, 43, 44, 45, 46, 47, 48, 8, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59, 5, 60, 61, 62, 63, 64, 10, 65, 66, 67, 11, 5, 68, 69, 70, 71, 72, 73, 74]]

```
word_to_index = {}
for i, sequence in enumerate(sequences):
  word_in_sentence = clean_sent[i].split()
    print(word_in_sentence)
   for j, value in enumerate(sequence):
        index_to_word[value] = word_in_sentence[j]
        word_to_index[word_in_sentence[j]] = value
print(index to word, "\n")
```

{2: 'deep', 1: 'learning', 12: 'also', 13: 'known', 6: 'as', 14: 'structured', 15: 'is', 16: 'part', 7: 'of', 17: 'a', 18: 'broader', 19: 'family', 8: 'm achine', 20: 'methods', 21: 'based', 22: 'on', 23: 'artificial', 4: 'neural', 3: 'networks', 24: 'with', 25: 'representation', 26: 'can', 27: 'be', 9: 's upervised', 28: 'semi', 29: 'or', 30: 'unsupervised', 31: 'architectures', 32: 'such', 33: 'belief', 34: 'reinforcement', 35: 'recurrent', 36: 'convoluti onal', 5: 'and', 37: 'transformers', 10: 'have', 38: 'been', 39: 'applied', 11: 'to', 40: 'fields', 41: 'including', 42: 'computer', 43: 'vision', 44: 's peech', 45: 'recognition', 46: 'natural', 47: 'language', 48: 'processing', 49: 'transformatics', 51: 'drug', 52: 'design', 53: 'medical', 54: 'image', 55: 'analysis', 56: 'climate', 57: 'science', 58: 'material', 59: 'inspection', 60: 'board', 61: 'game', 62: 'programs', 63: 'where', 64: 'they', 65: 'produced', 66: 'results', 67: 'comparable', 68: 'in', 69: 'some', 70: 'cases', 71: 'surpassing', 72: 'human', 73: 'expert', 74: 'performa nee')

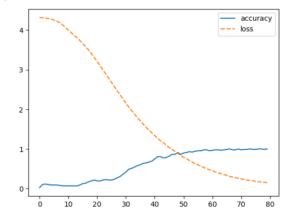
{'deep': 2, 'learning': 1, 'also': 12, 'known': 13, 'as': 6, 'structured': 14, 'is': 15, 'part': 16, 'of': 7, 'a': 17, 'broader': 18, 'family': 19, 'mach ine': 8, 'methods': 20, 'based': 21, 'on': 22, 'artificial': 23, 'neural': 4, 'networks': 3, 'with': 24, 'representation': 25, 'can': 26, 'be': 27, 'supe rvised': 9, 'semi': 28, 'or': 29, 'unsupervised': 30, 'architectures': 31, 'such': 32, 'belief': 33, 'reinforcement': 34, 'recurrent': 35, 'convolutiona 1': 36, 'and': 5, 'transformers': 37, 'have': 10, 'been': 38, 'applied': 39, 'to: 11, 'fields': 40, 'including': 41, 'computer': 42, 'vision': 43, 'spee ch': 44, 'recognition': 45, 'natural': 46, 'language': 47, 'processing': 48, 'translation': 49, 'bioinformatics': 50, 'drug': 51, 'design': 52, 'medica 1': 53, 'image': 54, 'analysis': 55, 'climate': 56, 'science': 57, 'material': 58, 'inspection': 59, 'board': 60, 'game': 61, 'programs': 62, 'where': 63, 'they': 64, 'produced': 65, 'results': 66, 'comparable': 67, 'in': 68, 'some': 69, 'cases': 70, 'surpassing': 71, 'human': 72, 'expert': 73, 'performa

```
[21]: vocab_size = len(tokenizer.word_index) + 1
                   context_size = 2
                   contexts = []
                   for sequence in sequences:
    for i in range(context_size, len(sequence) - context_size):
                                           \begin{array}{l} target = sequence[i] \\ context = [sequence[i-2], sequence[i-1], sequence[i+1], sequence[i+2]] \end{array} 
                                                 print(context)
                                            contexts.append(context)
                                           targets.append(target)
                    print(contexts, "\n")
                    print(targets)
                   [2, 1, 13, 6], [1, 12, 6, 2], [12, 13, 2, 14], [13, 6, 14, 1], [6, 2, 1, 15], [2, 14, 15, 16], [14, 1, 16, 7], [1, 15, 7, 17], [15, 16, 17, 18], [16, 7, 18, 19], [7, 17, 19, 7], [17, 18, 7, 8], [18, 19, 8, 1], [19, 7, 1, 20], [7, 8, 20, 21], [8, 1, 21, 22], [1, 20, 22, 23], [20, 21, 23, 4], [21, 22, 4, 3], [22, 23, 3, 24], [23, 4, 24, 25], [4, 3, 25, 1], [1, 26, 9, 28], [26, 27, 28, 9], [27, 9, 9, 29], [9, 28, 29, 30], [2, 1, 32, 6], [1, 31, 6, 2], [31, 32, 2, 4], [32, 6, 4, 3], [6, 2, 3, 2], [2, 4, 2, 33], [4, 3, 33, 3], [3, 2, 3, 2], [2, 33, 2, 34], [33, 3, 34, 1], [3, 2, 1, 35], [2, 34, 35, 4], [34, 1, 4, 3], [1, 35, 3, 36], [35, 4, 36, 4], [4, 3, 4, 3], [3, 2, 35], [36, 4, 5, 57], [4, 3, 37], 19], [31, 5, 10, 38], [5, 37, 38, 39], [37, 10, 39], [17, 10, 39], [17], [19, 38, 11, 40], [38, 39, 40, 41], [39, 11, 41, 42], [11, 40, 42, 43], [40, 41, 43, 44], [41, 42, 44, 45], [42, 43, 45, 46], [43, 44, 46, 47], [44, 45, 46], [43, 44, 46, 47], [44, 45], [45, 46, 48, 8], [46, 47, 8, 49], [47, 48, 49, 50], [48, 8, 50, 51], [8, 49, 51), [51, 52], [49, 50, 52, 53], [50, 51, 53, 54], [51, 52, 54, 55], [52, 53, 55], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [53, 54, 56], [54, 64, 66], [64, 66, 67
                   [12, 13, 6, 2, 14, 1, 15, 16, 7, 17, 18, 19, 7, 8, 1, 20, 21, 22, 23, 4, 3, 24, 27, 9, 28, 9, 31, 32, 6, 2, 4, 3, 2, 33, 3, 2, 34, 1, 35, 4, 3, 36, 4, 3, 5, 37, 10, 38, 39, 11, 40, 41, 42, 43, 44, 45, 46, 47, 48, 8, 49, 50, 51, 52, 53, 54, 55, 56, 57, 58, 59, 5, 60, 61, 62, 63, 64, 10, 65, 66, 67, 11, 5, 6
                    8, 69, 70, 71, 72]
[22]: #printing features with target
                    for i in range(5):
                               words = []
target = index_to_word.get(targets[i])
                               for j in contexts[i]:
                                            words.append(index_to_word.get(j))
                               print(words," -> ", target)
                   ['deep', 'learning', 'known', 'as'] -> also
['learning', 'also', 'as', 'deep'] -> known
['also', 'known', 'deep', 'structured'] -> as
['known', 'as', 'structured', 'learning'] -> de
['as', 'deep', 'learning', 'is'] -> structured
[23]: # Convert the contexts and targets to numpy arrays
                   X = np.array(contexts)
                   Y = np.array(targets)
[24]: # print(X)
[25]: import tensorflow as tf
                     from tensorflow.keras.models import Sequential
                     from tensorflow.keras.layers import Dense, Embedding, Lambda
[26]: model = Sequential([
                               Embedding(input_dim=vocab_size, output_dim=emb_size, input_length=2*context_size),
                               Lambda(lambda x: tf.reduce_mean(x, axis=1)),
                               Dense(256, activation='relu'),
Dense(512, activation='relu'),
                               Dense(vocab_size, activation='softmax')
                   C: \label{lem:condition} C: \label{lem:condi
                   ust remove it.
                     warnings.warn(
[27]: model.compile(loss='sparse_categorical_crossentropy', optimizer='adam', metrics=['accuracy'])
[28]: history = model.fit(X, Y, epochs=80)
                    Epoch 1/80
                                                                                        — 2s 14ms/step - accuracy: 0.0153 - loss: 4.3180
                     3/3 -
                    Epoch 2/80
                                                                             Os 7ms/step - accuracy: 0.1058 - loss: 4.3105
                    Epoch 3/80
                     3/3 -
                                                                                     — 0s 1ms/step - accuracy: 0.1428 - loss: 4.3024
                     Epoch 4/80
                                                                                          — 0s 10ms/step - accuracy: 0.1214 - loss: 4.2909
                    Epoch 5/80
                     3/3 -
                                                                                     --- Os 3ms/step - accuracy: 0.0845 - loss: 4.2765
                     Epoch 6/80

    Os 6ms/step - accuracy: 0.1001 - loss: 4.2567
```

```
Epoch 68/80
3/3
                        - 0s 7ms/step - accuracy: 0.9886 - loss: 0.2839
Epoch 69/80
3/3 -
                        - 0s 7ms/step - accuracy: 0.9808 - loss: 0.2774
Epoch 70/80
3/3 -
                        - 0s 5ms/step - accuracy: 1.0000 - loss: 0.2379
Epoch 71/80
3/3 -
                        - 0s 4ms/step - accuracy: 0.9808 - loss: 0.2463
Epoch 72/80
3/3 -
                        - 0s 4ms/step - accuracy: 0.9904 - loss: 0.2259
Epoch 73/80
3/3
                        - 0s 4ms/step - accuracy: 0.9826 - loss: 0.2120
Epoch 74/80
3/3 -
                        - 0s 4ms/step - accuracy: 1.0000 - loss: 0.2101
Epoch 75/80
3/3
                         Os 3ms/step - accuracy: 0.9943 - loss: 0.1889
Epoch 76/80
3/3
                        - 0s 3ms/step - accuracy: 0.9904 - loss: 0.1689
Epoch 77/80
                        - 0s 5ms/step - accuracy: 1.0000 - loss: 0.1769
3/3 -
Epoch 78/80
                        - 0s 4ms/step - accuracy: 1.0000 - loss: 0.1737
Epoch 79/80
                        - 0s 2ms/step - accuracy: 0.9904 - loss: 0.1636
3/3 -
Epoch 80/80
3/3
                        - 0s 4ms/step - accuracy: 1.0000 - loss: 0.1634
```

```
[29]: <Axes: >
```



```
[30]: # from sklearn.decomposition import PCA

# embeddings = model.get_weights()[0]

# pca = PCA(n_components=2)

# reduced_embeddings = pca.fit_transform(embeddings)
```

[31]: print("'Deep learning (also known as deep structured learning) is part of a broader family of machine learning methods based on artificial neural networks

Deep learning (also known as deep structured learning) is part of a broader family of machine learning methods based on artificial neural networks with representation learning. Learning can be supervised, semi-supervised or unsupervised. Deep-learning architectures such as deep neural networks, deep beli ef networks, deep reinforcement learning, recurrent neural networks, convolutional neural networks and Transformers have been applied to fields including computer vision, speech recognition, natural language processing, machine translation, bioinformatics, drug design, medical image analysis, climate scien ce, material inspection and board game programs, where they have produced results comparable to and in some cases surpassing human expert performance.

```
[32]: # test model: select some sentences from above paragraph
test_sentenses = [
    "known as structured learning",
    "transformers have applied to",
    "where they produced results",
    "cases surpassing expert performance",
    "comparable to in some",
    "some cases surpassing human"
]
```