

## Exercise 2.2: Complex machine learning models

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### Model: CNN

I chose CNN for this analysis because although the data involves a time component, the current analysis disregards that component. CNN is more appropriate because it can handle the complexity of the data.

Version 1:

### Starting Hyperparameters:

```
epochs = 5
batch_size = 16
n_hidden = 4

timesteps = len(X_train[0])
input_dim = len(X_train[0][0])
n_classes = len(y_train[0])

model = Sequential()
model.add(Conv1D(n_hidden, kernel_size=2, activation='relu', input_shape=(timesteps, input_dim)))
model.add(Dense(16, activation='relu'))
model.add(MaxPooling1D())
model.add(Flatten())
model.add(Dense(n_classes, activation='softmax')) # Options: sigmoid, tanh, softmax, relu
```

### Initial Confusion Matrix:

144/144		0s 2ms/step							
Pred		DEBILT	HEATHROW	KASSEL	LJUBLJANA	MAASTRICHT	MADRID	OSLO	\
True									
BASEL		102	79	9	25	2	239	5	
BELGRADE		0	0	0	0	4	9	0	
BUDAPEST		0	0	0	0	0	2	0	
DEBILT		0	0	0	0	0	0	0	
DUSSELDORF		0	0	0	0	0	0	0	
HEATHROW		0	0	0	0	0	1	0	
KASSEL		0	0	0	0	0	0	0	
LJUBLJANA		0	0	0	0	0	0	0	
MAASTRICHT		0	0	0	0	0	0	0	
MADRID		0	0	0	0	0	27	0	
MUNCHENB		0	0	0	0	0	0	0	
OSLO		0	0	0	0	0	0	0	
STOCKHOLM		0	0	0	0	0	0	0	
VALENTIA		0	0	0	0	0	0	0	
Pred	STOCKHOLM								
True									
BASEL		2507							
BELGRADE		844							
BUDAPEST		191							
DEBILT		72							
DUSSELDORF		25							
HEATHROW		69							
KASSEL		7							
LJUBLJANA		30							
MAASTRICHT		8							
MADRID		321							
MUNCHENB		4							
OSLO		6							
STOCKHOLM		1							
VALENTIA		1							

8 stations, accuracy 11%, loss 785,503

Version 2:

Hyperparameters:

```
epochs = 20
batch_size = 25
n_hidden = 8

timesteps = len(X_train[0])
input_dim = len(X_train[0][0])
n_classes = len(y_train[0])

model = Sequential()
model.add(Conv1D(n_hidden, kernel_size=2, activation='relu', input_shape=(timesteps, input_dim)))
model.add(Dense(16, activation='relu'))
model.add(MaxPooling1D())
model.add(Flatten())
model.add(Dense(n_classes, activation='softmax')) # Options: sigmoid, tanh, softmax, relu
```

Confusion Matrix:

144/144		1s 8ms/step						
Pred		BASEL	BELGRADE	BUDAPEST	DEBILT	HEATHROW	KASSEL	LJUBLJANA \
True								
BASEL		143	82	158	107	38	40	453
BELGRADE		0	3	132	4	0	0	137
BUDAPEST		1	0	25	6	0	0	28
DEBILT		1	0	2	3	0	0	5
DUSSELDORF		0	0	1	0	0	0	2
HEATHROW		0	0	9	1	0	0	11
KASSEL		0	0	0	0	0	0	1
LJUBLJANA		0	0	2	0	0	0	19
MAASTRICHT		0	0	0	0	0	0	1
MADRID		25	1	63	17	0	0	82
MUNCHENB		0	0	2	0	0	0	2
OSLO		0	0	1	0	0	0	1
STOCKHOLM		0	0	0	0	0	0	1
VALENTIA		0	0	0	0	0	0	0

Pred		MAASTRICHT	MUNCHENB	OSLO	SONNBLICK	STOCKHOLM	VALENTIA
True							
BASEL		11	1844	42	6	18	26
BELGRADE		0	581	0	0	0	0
BUDAPEST		0	133	0	0	0	0
DEBILT		0	61	0	0	0	0
DUSSELDORF		0	22	0	0	0	0
HEATHROW		0	49	0	0	0	0
KASSEL		0	6	0	0	0	0
LJUBLJANA		0	9	0	0	0	0
MAASTRICHT		0	7	0	0	0	0
MADRID		0	156	2	0	2	0
MUNCHENB		0	0	0	0	0	0
OSLO		0	4	0	0	0	0
STOCKHOLM		0	0	0	0	0	0
VALENTIA		0	1	0	0	0	0

13 stations, accuracy 12.8%, loss 2,165,907

Version 3:

Hyperparameters:

```
epochs = 30
batch_size = 25
n_hidden = 16

timesteps = len(X_train[0])
input_dim = len(X_train[0][0])
n_classes = len(y_train[0])

model = Sequential()
model.add(Conv1D(n_hidden, kernel_size=2, activation='relu', input_shape=(timesteps, input_dim)))
model.add(Dense(16, activation='relu'))
model.add(MaxPooling1D())
model.add(Flatten())
model.add(Dense(n_classes, activation='softmax')) # Options: sigmoid, tanh, softmax, relu
```

Confusion matrix:

144/144		1s 6ms/step						
Pred		BASEL	BELGRADE	BUDAPEST	DEBILT	DUSSELDORF	KASSEL	LJUBLJANA \
True								
BASEL		1458	49	85	334	22	3	311
BELGRADE		286	0	3	25	0	0	217
BUDAPEST		73	0	1	12	0	0	39
DEBILT		31	0	0	10	0	0	13
DUSSELDORF		13	0	0	6	0	0	2
HEATHROW		32	0	0	8	0	0	2
KASSEL		4	0	0	1	0	0	1
LJUBLJANA		13	0	0	2	0	0	5
MAASTRICHT		8	0	0	0	0	0	0
MADRID		164	0	0	46	0	0	40
MUNCHENB		0	0	0	0	0	0	4
OSLO		0	0	0	1	0	0	1
STOCKHOLM		0	0	0	0	0	0	1
VALENTIA		1	0	0	0	0	0	0

Pred		MADRID	MUNCHENB	OSLO	SONNBLICK	STOCKHOLM	VALENTIA
True							
BASEL		208	413	38	6	12	29
BELGRADE		135	190	1	0	0	0
BUDAPEST		41	26	1	0	0	0
DEBILT		10	7	1	0	0	0
DUSSELDORF		3	1	0	0	0	0
HEATHROW		26	2	0	0	0	0
KASSEL		0	1	0	0	0	0
LJUBLJANA		10	0	0	0	0	0
MAASTRICHT		0	0	0	0	0	0
MADRID		87	2	7	0	2	0
MUNCHENB		0	0	0	0	0	0
OSLO		3	1	0	0	0	0
STOCKHOLM		0	0	0	0	0	0
VALENTIA		0	0	0	0	0	0

13 stations, accuracy 11.6%, loss 10,374,783

*Loss is increasing exponentially. We will need to change the activation type for v4.*

Version 4:

Hyperparameters:

```
epochs = 30
batch_size = 30
n_hidden = 64

timesteps = len(X_train[0])
input_dim = len(X_train[0][0])
n_classes = len(y_train[0])

model = Sequential()
model.add(Conv1D(n_hidden, kernel_size=2, activation='sigmoid', input_shape=(timesteps, input_dim)))
model.add(Dense(16, activation='sigmoid'))
model.add(MaxPooling1D())
model.add(Flatten())
model.add(Dense(n_classes, activation='relu')) # Options: sigmoid, tanh, softmax, relu
```

Confusion matrix:

**144/144**  **1s 6ms/step**

Pred	BASEL
True	
BASEL	2968
BELGRADE	857
BUDAPEST	193
DEBILT	72
DUSSELDORF	25
HEATHROW	70
KASSEL	7
LJUBLJANA	30
MAASTRICHT	8
MADRID	348
MUNCHENB	4
OSLO	6
STOCKHOLM	1
VALENTIA	1

*Well, that's not going to work. Accuracy is 64% and loss is nan, but it only found one station. Let's try one more time.*

## Version 4

### Hyperparameters:

```
epochs = 30
batch_size = 16
n_hidden = 256

timesteps = len(X_train[0])
input_dim = len(X_train[0][0])
n_classes = len(y_train[0])

model = Sequential()
model.add(Conv1D(n_hidden, kernel_size=2, activation='tanh', input_shape=(timesteps, input_dim)))
model.add(Dense(16, activation='tanh'))
model.add(MaxPooling1D())
model.add(Flatten())
model.add(Dense(n_classes, activation='softmax')) # Options: sigmoid, tanh, softmax, relu
```

### Confusion Matrix:

144/144		1s 8ms/step						
Pred	BASEL	BELGRADE	BUDAPEST	DEBILT	DUSSELDORF	HEATHROW	KASSEL	\
True								
BASEL	3	204	39	15	39	48	141	
BELGRADE	0	118	19	0	0	0	0	
BUDAPEST	0	10	5	0	0	0	0	
DEBILT	0	1	3	0	0	0	0	
DUSSELDORF	0	0	0	0	0	0	0	
HEATHROW	0	0	0	1	0	1	0	
KASSEL	0	2	0	0	0	0	0	
LJUBLJANA	0	3	0	0	0	0	0	
MAASTRICHT	0	0	0	0	0	0	0	
MADRID	0	3	0	0	0	1	0	
MUNCHENB	0	0	1	0	0	0	0	
OSLO	0	0	0	0	0	0	0	
STOCKHOLM	0	1	0	0	0	0	0	
VALENTIA	0	0	0	0	0	0	0	
Pred	LJUBLJANA	MAASTRICHT	MADRID	MUNCHENB	OSLO	SONNBLICK	\	
True								
BASEL	15	24	1365	2	87	63		
BELGRADE	3	0	716	1	0	0		
BUDAPEST	1	0	177	0	0	0		
DEBILT	0	0	68	0	0	0		
DUSSELDORF	0	0	25	0	0	0		
HEATHROW	0	0	68	0	0	0		
KASSEL	0	0	5	0	0	0		
LJUBLJANA	0	0	27	0	0	0		
MAASTRICHT	0	0	7	1	0	0		
MADRID	0	0	342	0	0	2		
MUNCHENB	0	0	2	1	0	0		
OSLO	0	0	6	0	0	0		
STOCKHOLM	0	0	0	0	0	0		
VALENTIA	0	0	1	0	0	0		
Pred	STOCKHOLM	VALENTIA						
True								
BASEL	102	821						
BELGRADE	0	0						
BUDAPEST	0	0						
DEBILT	0	0						
DUSSELDORF	0	0						
HEATHROW	0	0						
KASSEL	0	0						
LJUBLJANA	0	0						
MAASTRICHT	0	0						
MADRID	0	0						
MUNCHENB	0	0						
OSLO	0	0						
STOCKHOLM	0	0						
VALENTIA	0	0						

The model now recognizes all 15 stations, with 10.4% accuracy and loss of 40.7