

Metagenomic Classification with Deep Learning: Experiments #1

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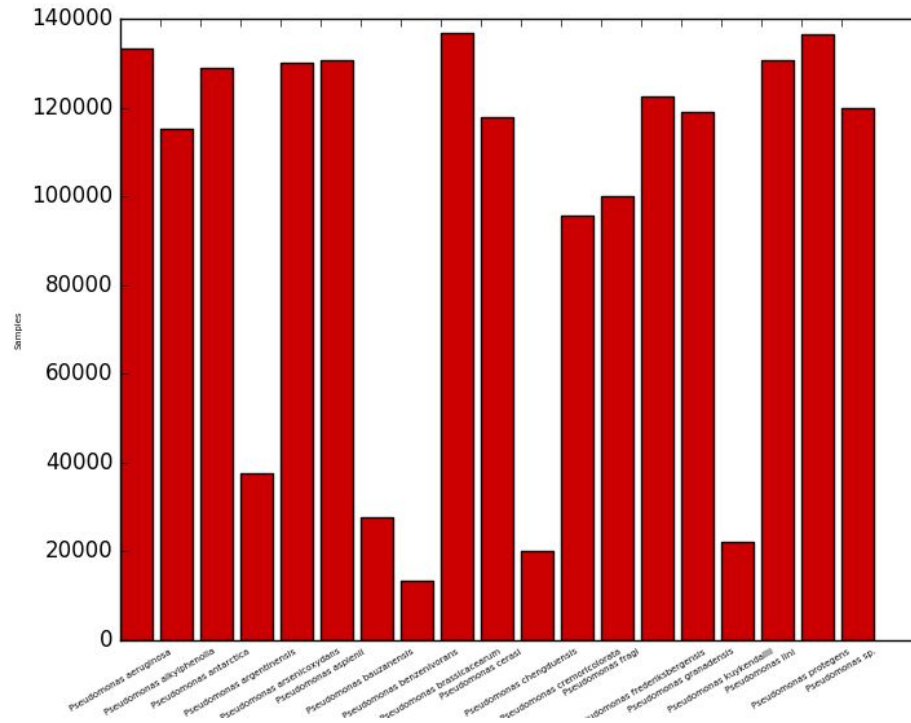
Goal

Recognize a specie among others:
one-against-all binary classification

Dataset

- 19 genomes belonging to species of the genus *Pseudomonas*
- Minimum of 90% similarity among them
- Splitting of each genome in 250 base pair **reads**
 - Shifting a cropping window by 50 bp
- Total number of reads per genome = $((\text{GenomeLength}-250)/50)+1$
- Max number of reads: *Pseudomonas brassicacearum*, 136860
- Min number of reads: *Pseudomonas benzenivorans*, 13459

Dataset reads distribution



**Test #1: *Pseudomonas*
benzenivorans against all
(least-number-of-reads specie)**

Test 1.1 - Goal

- Study the performances of a chosen network on different training sets
- The training sets differ by composition, i.e. positive/negative examples proportion

Dataset preparation

- **Training set:**
 - 80% of positive class reads ($\text{Pos} = 10767$)
 - A number of negative examples proportional to Pos:
 - $\text{Neg} = \text{Pos} \times 8, \text{Pos} \times 12, \text{Pos} \times 16, \text{Pos} \times 18, \text{Pos} \times 20, \text{Pos} \times 24$
- **Test set:**
 - 20% of positive class reads ($\text{Pos} = 2692$)
 - Same amount from all other classes ($\text{Neg} = \text{Pos} \times 18 = 51130$)

Neural network

1. Convolutional(#kernels = 64, kernel_size = 28, activation = ReLU)
2. Convolutional(#kernels = 64, kernel_size = 5, activation = ReLU)
3. MaxPooling(kernel_size = 2)
4. Convolutional(#kernels = 64, kernel_size = 3, activation = ReLU)
5. GlobalMaxPooling(kernel_size = 2)
6. Dense(#neurons = 2, activation = Softmax)

- Learning rate: 0.0005
- Optimizer: Adam
- Epochs: 21
- Batch size: 250

Neural network

Layer (type)	Output Shape	Param #
conv1d_4 (Conv1D)	(None, 223, 64)	9024
conv1d_5 (Conv1D)	(None, 219, 64)	20544
max_pooling1d_2 (MaxPooling1	(None, 109, 64)	0
conv1d_6 (Conv1D)	(None, 107, 64)	12352
global_max_pooling1d_2 (Glob	(None, 64)	0
dense_2 (Dense)	(None, 2)	130

Total params: 42,050

Trainable params: 42,050

Non-trainable params: 0

Training set Neg = Pos*8

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2112	580
	Neg	1039	47399

- Precision: 0.670
- Recall: 0.784
- F1: 0.722

Training set Neg = Pos*12

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2392	300
	Neg	1042	47396

- Precision: 0.696
- Recall: 0.888
- F1: 0.780

Training set Neg = Pos*16

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	1939	753
	Neg	632	47806

- Precision: 0.762
- Recall: 0.845
- F1: 0.801

Training set Neg = Pos*18

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	1712	980
	Neg	182	48256

- Precision: 0.903
- Recall: 0.635
- F1: 0.746

Training set Neg = Pos*20

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2140	552
	Neg	557	47881

- Precision: 0.793
- Recall: 0.794
- F1: 0.794

Training set Neg = Pos*24

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2373	319
	Neg	714	47724

- Precision: 0.768
- Recall: 0.881
- F1: 0.821

Test 1.2 - Goal

- Changing some of the parameters of the network, in order to improve performances.
- Training set chosen as the one which brought the worst recall ($\text{Neg} = 18 * \text{Pos}$).

Neural network

1. Convolutional(#kernels = 64, kernel_size = 28, activation = ReLU)
 2. Convolutional(#kernels = 96, kernel_size = 5, activation = ReLU)
 3. MaxPooling(kernel_size = 2)
 4. Convolutional(#kernels = 96, kernel_size = 3, activation = ReLU)
 5. GlobalMaxPooling(kernel_size = 2)
 6. Dense(#neurons = 2, activation = Softmax)
- Learning rate: 0.0005
 - Optimizer: Adam
 - Epochs: 21
 - Batch size: 250

Kernels: 64-96-96

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2466	226
	Neg	398	48040

- Precision: 0.861
- Recall: 0.916
- F1: 0.887

Neural network

1. Convolutional(#kernels = 96, kernel_size = 28, activation = ReLU)
 2. Convolutional(#kernels = 96, kernel_size = 5, activation = ReLU)
 3. MaxPooling(kernel_size = 2)
 4. Convolutional(#kernels = 96, kernel_size = 3, activation = ReLU)
 5. GlobalMaxPooling(kernel_size = 2)
 6. Dense(#neurons = 2, activation = Softmax)
- Learning rate: 0.0005
 - Optimizer: Adam
 - Epochs: 21
 - Batch size: 250

Kernels: 96-96-96

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2442	250
	Neg	269	48169

- Precision: 0.900
- Recall: 0.907
- F1: 0.903

Neural network

1. Convolutional(#kernels = 96, kernel_size = 21, activation = ReLU)
2. Convolutional(#kernels = 96, kernel_size = 5, activation = ReLU)
3. MaxPooling(kernel_size = 2)
4. Convolutional(#kernels = 96, kernel_size = 3, activation = ReLU)
5. GlobalMaxPooling(kernel_size = 2)
6. Dense(#neurons = 2, activation = Softmax)

- Learning rate: 0.0005
- Optimizer: Adam
- Epochs: 21
- Batch size: 250

Kernels: 96-96-96, Conv1 kernel_size 21

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2511	181
	Neg	290	48148

- Precision: 0.896
- Recall: 0.932
- F1: 0.914

Neural network

1. Convolutional(#kernels = 96, kernel_size = 35, activation = ReLU)
2. Convolutional(#kernels = 96, kernel_size = 5, activation = ReLU)
3. MaxPooling(kernel_size = 2)
4. Convolutional(#kernels = 96, kernel_size = 3, activation = ReLU)
5. GlobalMaxPooling(kernel_size = 2)
6. Dense(#neurons = 2, activation = Softmax)

- Learning rate: 0.0005
- Optimizer: Adam
- Epochs: 21
- Batch size: 250

Kernels: 96-96-96, Conv1 kernel_size 35

- Confusion matrix:

		Predicted	
		Pos	Neg
Real	Pos	2421	271
	Neg	179	48259

- Precision: 0.931
- Recall: 0.899
- F1: 0.914