

# Supplementary Material

Identification of Near-Fault Impulsive Signals and Their  
Initiation and Termination Positions with Convolutional  
Neural Networks

Deniz Ertuncay<sup>1</sup>, Andrea de Lorenzo<sup>2</sup>, Giovanni Costa<sup>1</sup>

<sup>1</sup> University of Trieste, Department of Mathematics and  
Geosciences,  
SeisRaM Working Group, Via Eduardo Weiss 4, 34128 Trieste,  
Italy

<sup>2</sup> University of Trieste, Department of Engineering and  
Architecture,  
Machine Learning Lab, Via Valerio, 7/4, 34127 Trieste, Italy

# Identification of Impulsive Signals

Confusion matrices of methods for  $k^2$  and Mavroeidis models for each fold are presented in the supplementary material.

## $k^2$ Model

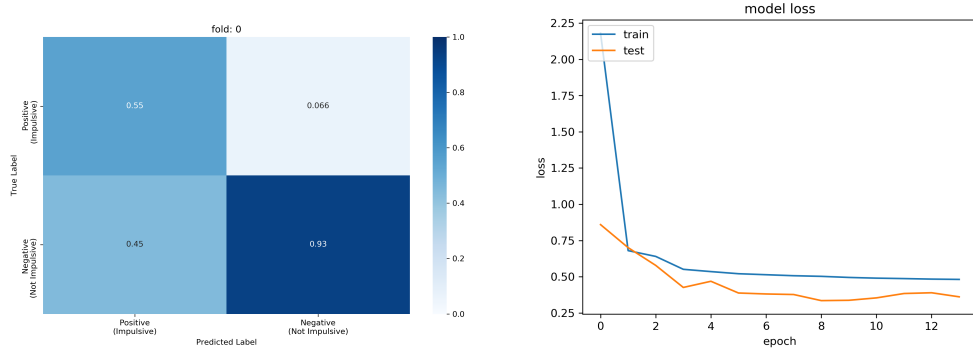


Figure S1: Confusion matrix (left) and loss function (right) of Fold 0

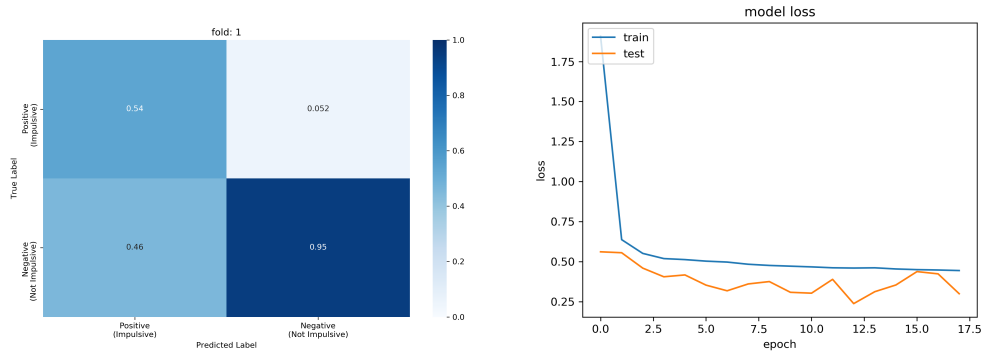


Figure S2: Confusion matrix (left) and loss function (right) of Fold 1

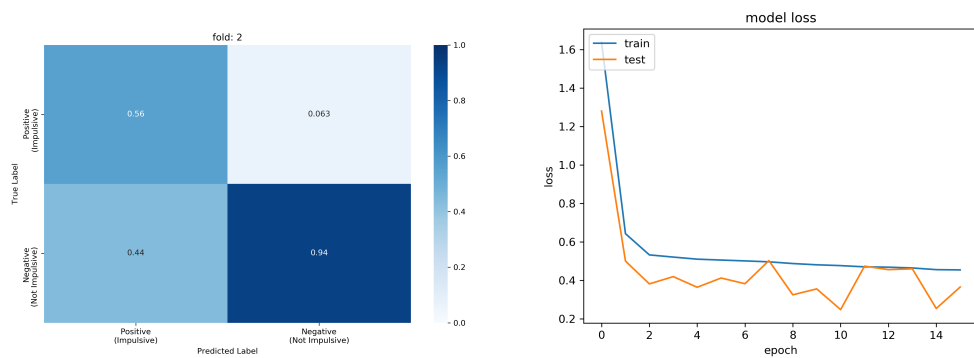


Figure S3: Confusion matrix (left) and loss function (right) of Fold 2

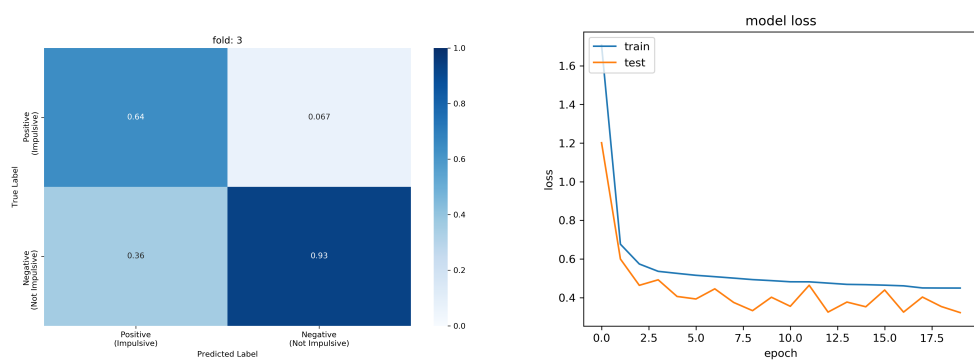


Figure S4: Confusion matrix (left) and loss function (right) of Fold 3

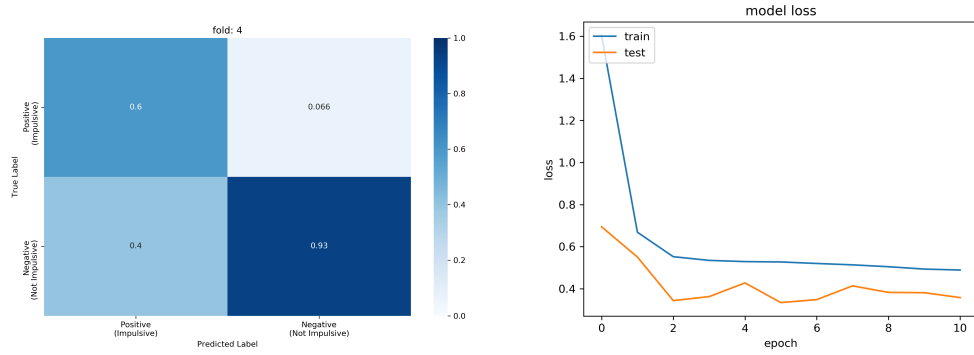


Figure S5: Confusion matrix (left) and loss function (right) of Fold 4

## Mavroeidis Model

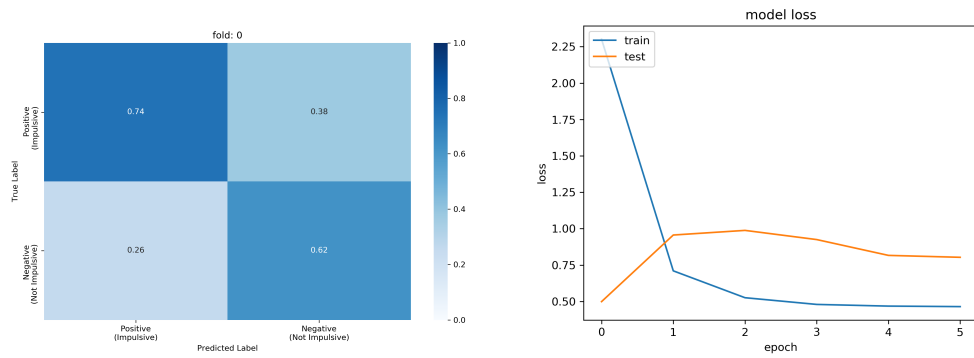


Figure S6: Confusion matrix (left) and loss function (right) of Fold 0

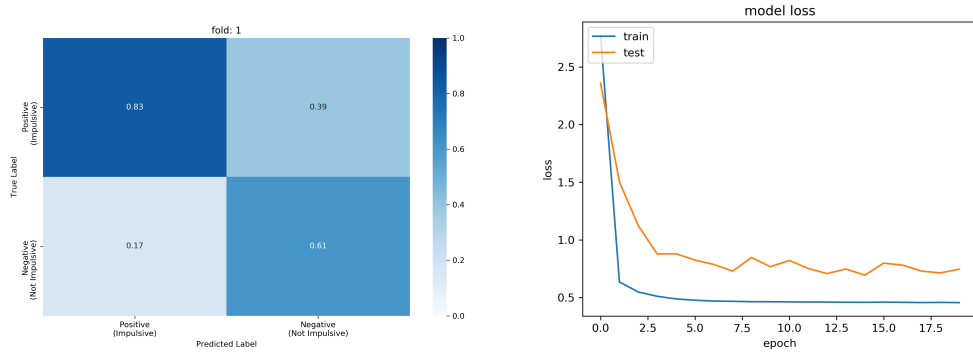


Figure S7: Confusion matrix (left) and loss function (right) of Fold 1

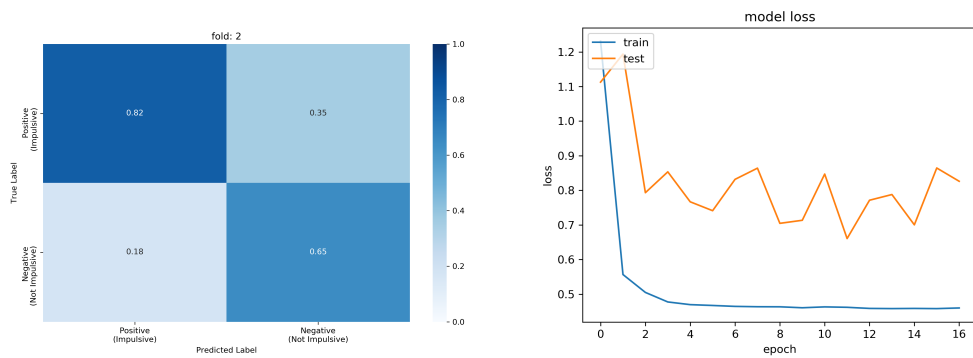


Figure S8: Confusion matrix (left) and loss function (right) of Fold 2

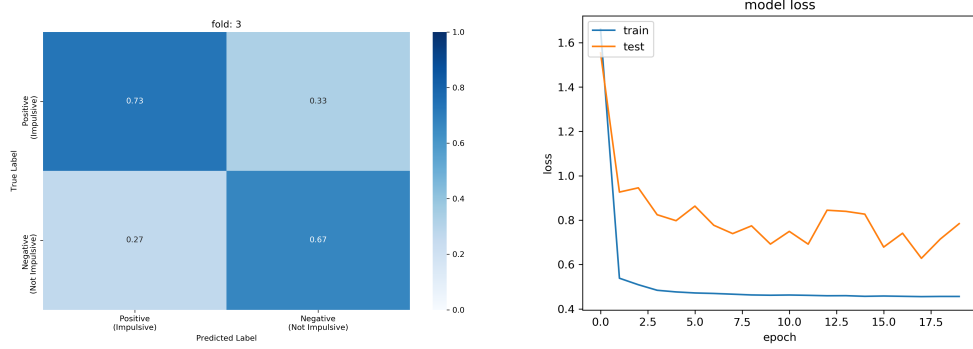


Figure S9: Confusion matrix (left) and loss function (right) of Fold 3

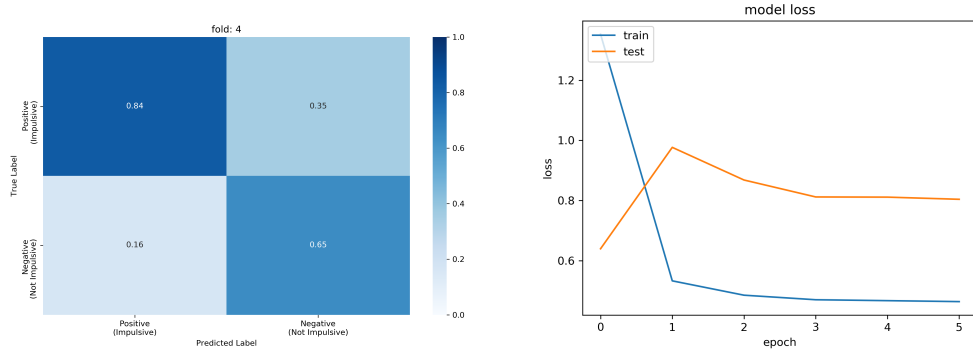


Figure S10: Confusion matrix (left) and loss function (right) of Fold 4

## Determination of Initiation and Termination Positions of Impulsive Signals

In order to assess the architecture of our proposal, we repeated all the experiments removing some components of the neural network. In particular, we perform an experiment removing the additional input  $\arg \max(\vec{w})$  and  $\arg \min(\vec{w})$  for evaluating the contribution of these two elements (“NO-RARG” experiment). As reported in Table 1, results are always slightly worse without these two inputs. Moreover, we investigated whether the contribution of these two inputs alone was sufficient to estimate initiation and

termination positions. We performed an experiment with a very simple fully connected neural network (without the convolutional part) which takes as input only the vector  $\langle \max(\vec{w}), \min(\vec{w}), \arg \max(\vec{w}), \arg \min(\vec{w}) \rangle$  (“NOCONV” experiment). Despite the results being surprisingly good, they are always worse than compared with the CNN, which confirms that the whole architecture is necessary to achieve better results.

Table S1: Comparison of all the method variations averaged among 5-fold cross-validation

Method	$R^2$			MAE			MSE		
	$s$	$e$	$\mu$	$s$	$e$	$\mu$	$s$	$e$	$\mu$
CNN	<b>0.97</b>	<b>0.97</b>	<b>0.97</b>	<b>17.51</b>	<b>22.53</b>	<b>20.03</b>	<b>610.23</b>	<b>1057.47</b>	<b>833.85</b>
NOARG	0.82	0.92	0.87	25.05	24.77	24.91	1290.32	1063.34	1176.83
NOCONV	0.83	0.86	0.85	21.67	28.29	24.98	1017.89	1782.13	1400.01