



FORTISSIMO



# Hyper-parametric search using HPC infrastructures for Tensorflow.

Gonzalo Ferro - CESGA.

Fortissimo experiment 707: Cyber-Physical Laser Metal Deposition (CyPLAM)

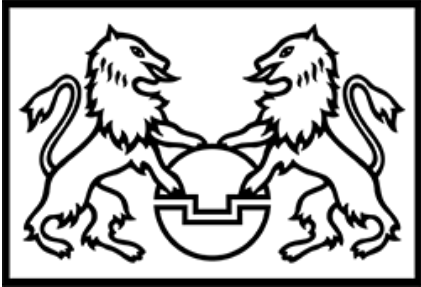


Co-funded by  
the European Union



This project has received funding from the European Union's H2020 research and innovation programme under grant agreement No 680481

# Experiment Partners:



End User: EMO

Die and mould manufacturer (Slovenia)



Domain Expert & ISV: AIMEN

Expert in laser technologies (Vigo, Spain)



HPC Centre: CESGA

HPC expert and provider (Santiago de Compostela, Spain)

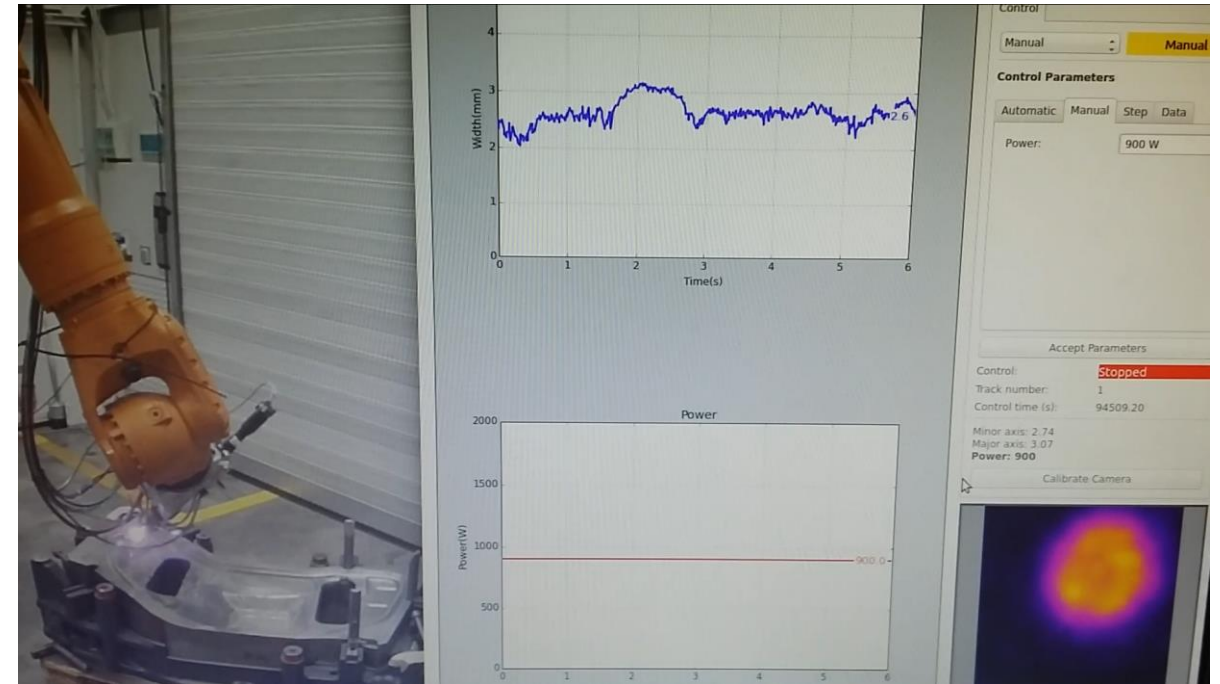


# Experiment Description:

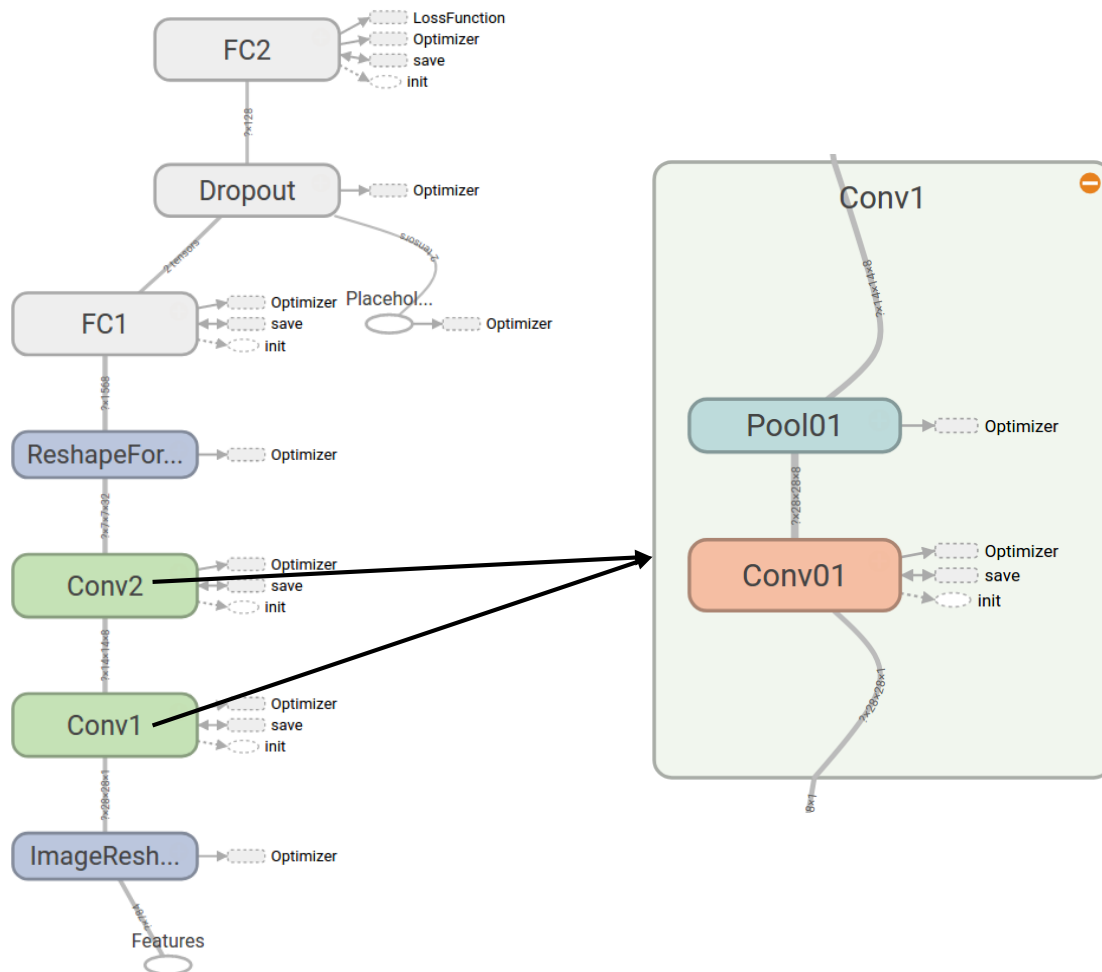
Laser Metal Deposition (LMD) for building and repairing large metal parts.

LMD process recorder by Medium Wavelength Infrared (MWIR) sensors attached to laser header.

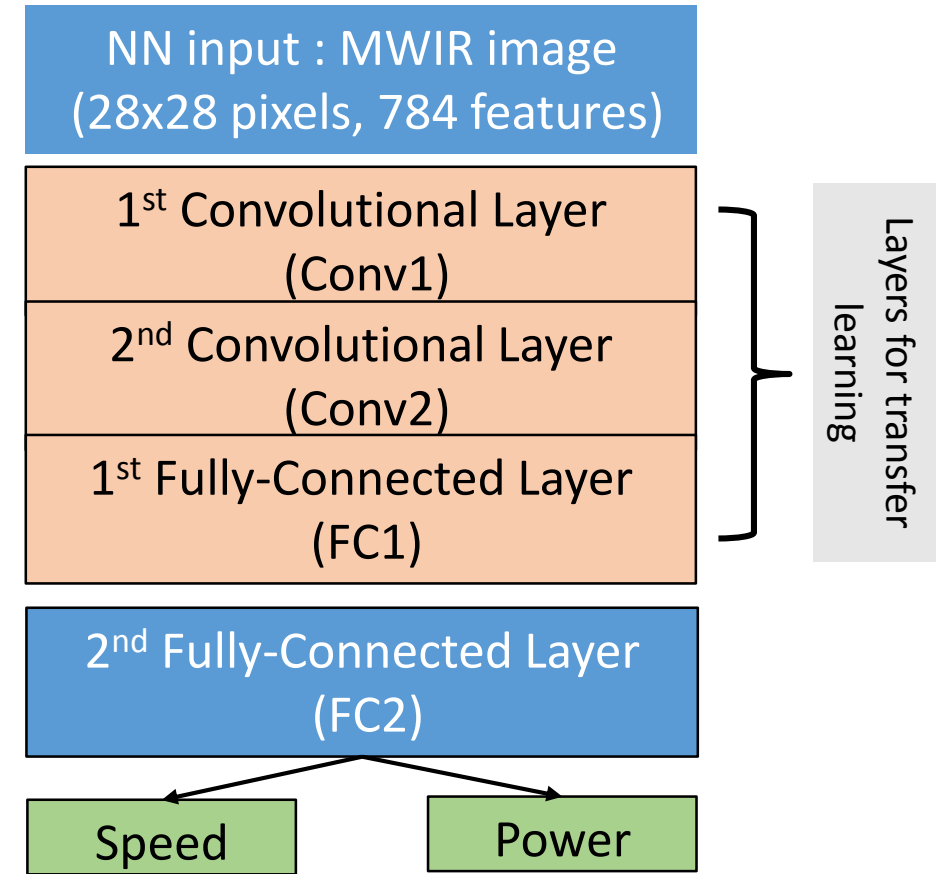
Use Convolutional Neural Network model for monitoring the LMD process based on the MWIR images.



# Model Description



NN Graph model (extracted from Tensorboard)



# Experiment Description: Loss Function

$$Loss = \frac{1}{2m} \sum_{i=1}^m \left[ (h_{LP}(x^i) - y_{LP}^i)^2 + (h_{LS}(x^i) - y_{LS}^i)^2 \right]$$

- $x^i$  is the  $i^{\text{th}}$  frame of the training dataset.
- $h_{LP}(x^i)$  is the prediction for Laser Power of the NN model for  $i^{\text{th}}$  frame.
- $y_{LP}^i$  : label for Laser Power of the  $i^{\text{th}}$  frame.
- $h_{LS}(x^i)$  is the prediction for Laser Speed of the NN model for  $i^{\text{th}}$  frame.
- $y_{LS}^i$  : label for Laser Speed of the  $i^{\text{th}}$  frame.
- $m$  : total number of training frames

# Experiment Description: Metrics

$$Err^i(x^i, y_{LP}^i, y_{LS}^i) = \text{Max} \left( \frac{y_{LP}^i}{h_{LP}(x^i)} - 1, \frac{y_{LS}^i}{h_{LS}(x^i)} - 1 \right)$$

$$Accuracy_{20\%} = \frac{1}{n} \sum_{i=1}^n (Err^i(x^i, y_{LP}^i, y_{LS}^i) < 20\%)$$

# Experiment Description: Dataset

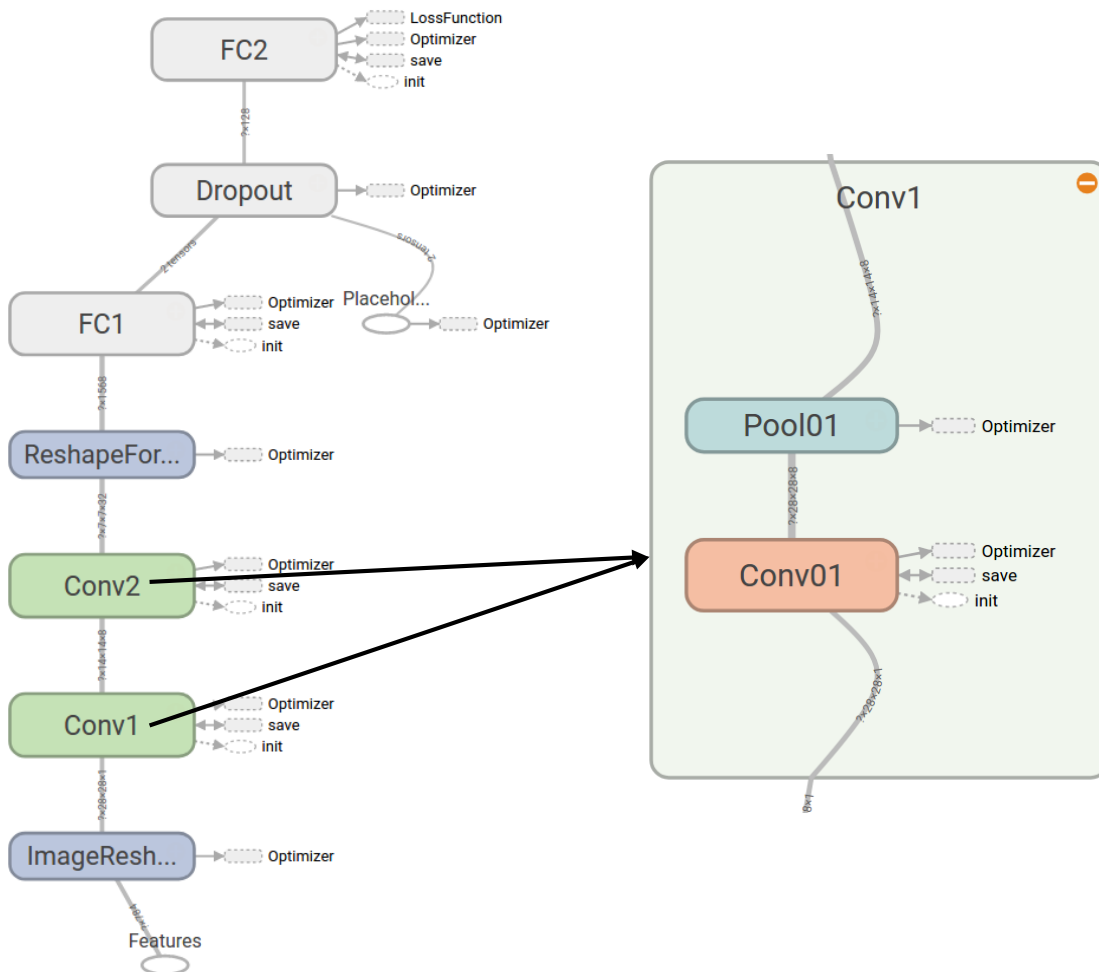
414 LMD tracks stainless steel (316L), powder & base material.

324 tracks for training

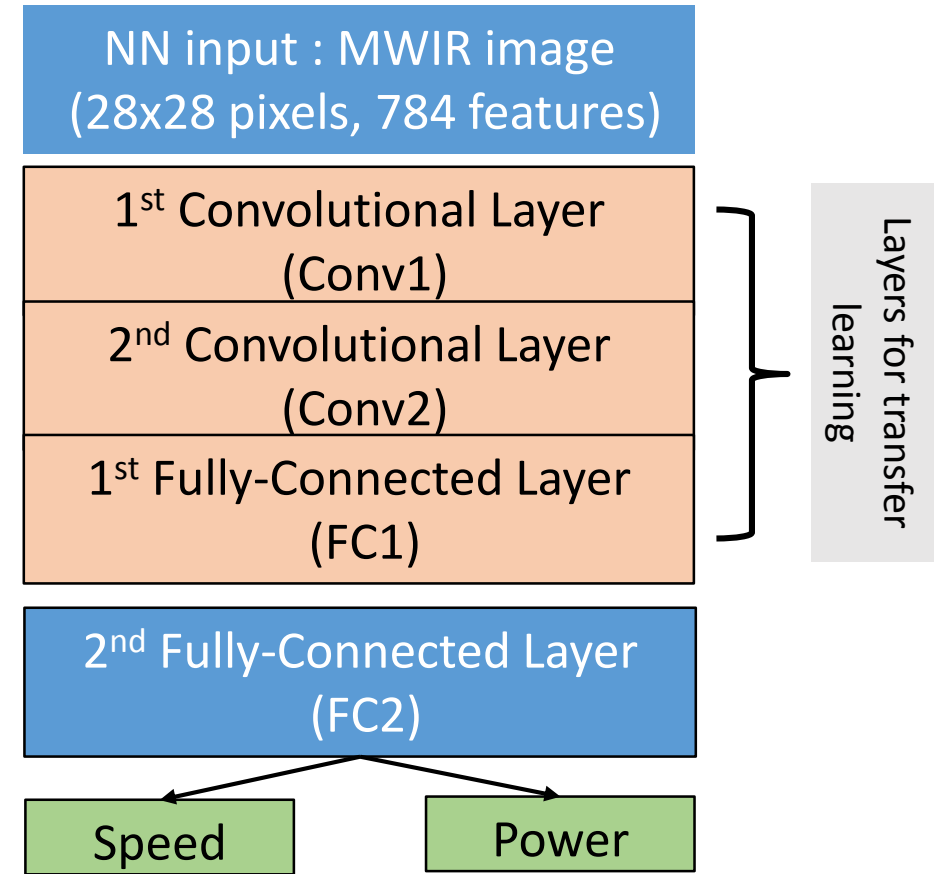
90 tracks for validation

Dataset	Number of frames	Pixels	Number of labels
Training	44,505	28 x 28 = 784	2
Validation	162,216	28 x 28 = 784	2

# Model Description

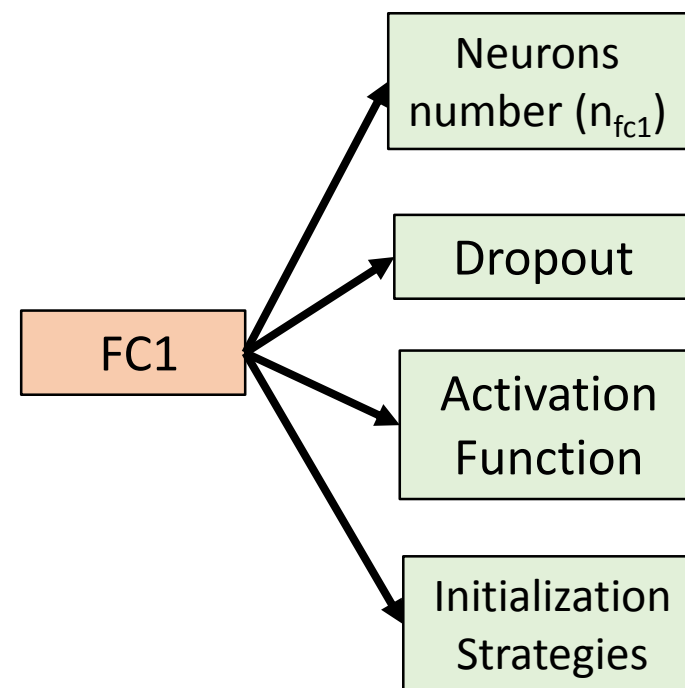
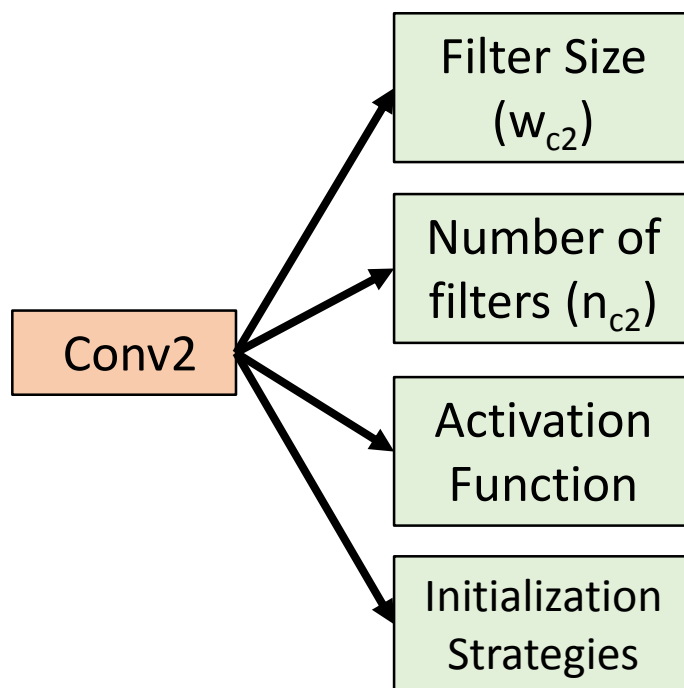
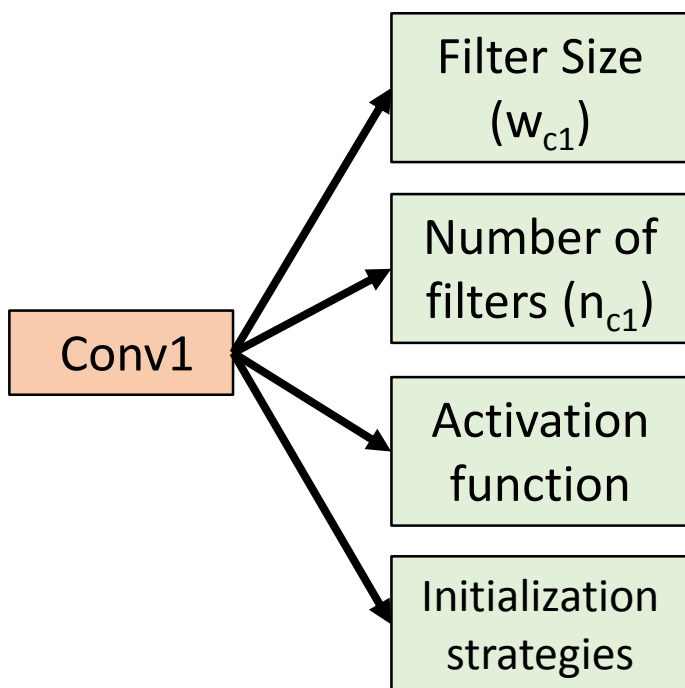


NN Graph model (extracted from Tensorboard)

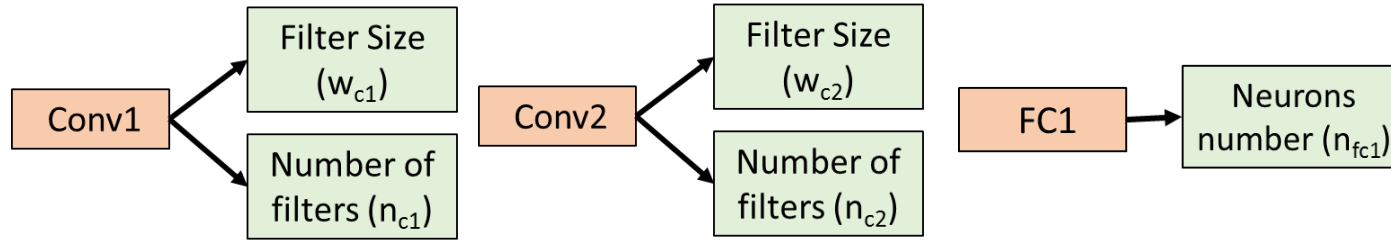




# Hyper Parametric Search: Model Parameters



# Hyper Parametric Search: Implementation



	Filter Size	Number of Filters
Convolutional Layer 01	$w_{c1}=3,5$	$n_{cv1}= 8, 16, 32$
Convolutional Layer 02	$w_{c2}=3,5$	$n_{cv2}= 32, 64$
	Number of Neurons	
Fully-connected Layer	$n_{fc1}=128, 256, 384, 512, 640, 768, 896, 1024$	

Hyper Parameter	Value
Learning Rate:	$10^{-5}$
Dropout:	0.75
Batch size:	128
Optimizer:	Adam
Regularizers:	None
Iterations	400k

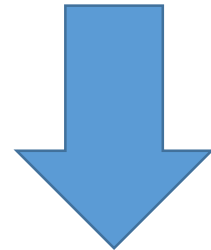
Grid search strategy

192 possible models!!!

200k-3M free parameters

# Results: time performance

192 Models: 40 days



factor 13.7

70 hours

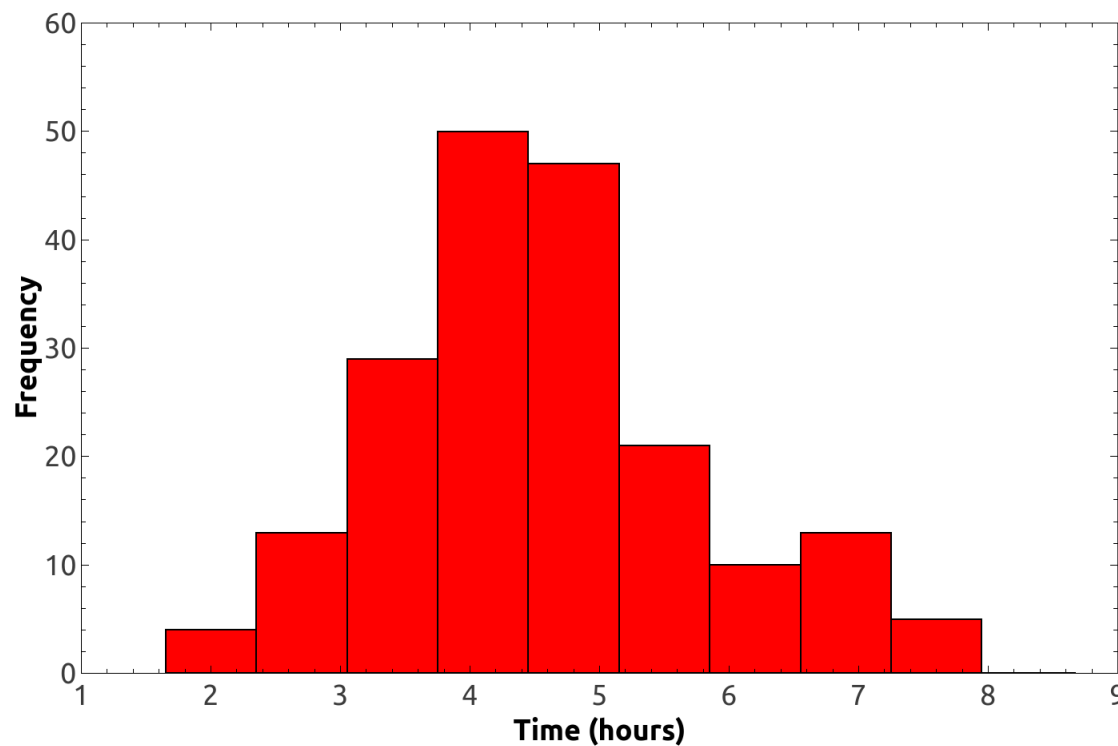
# Results: time performance

192 Models: 40 days  70 hours

Average trainings: 4-5 hours

Shortest training: 2.4 hours

Largest training: 8.2 hours



13.7



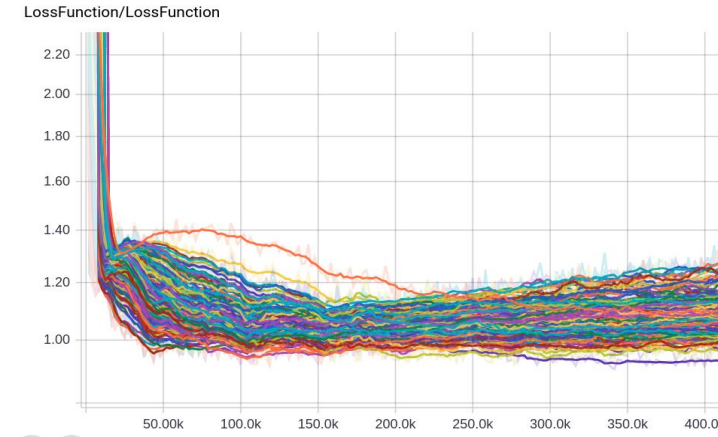
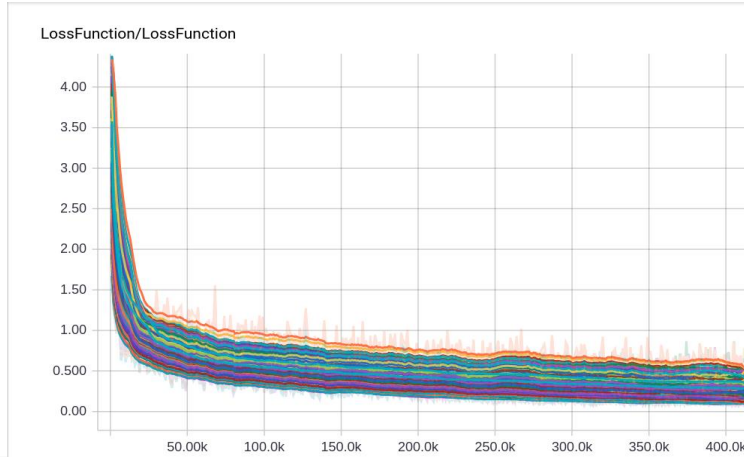
100 !!!!

# Results: Overall Performance

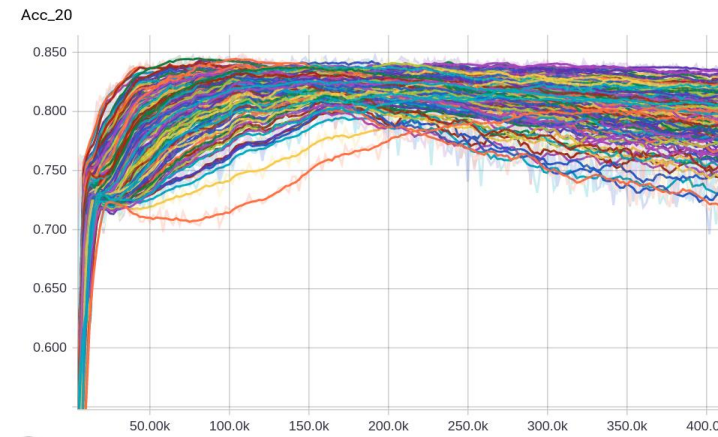
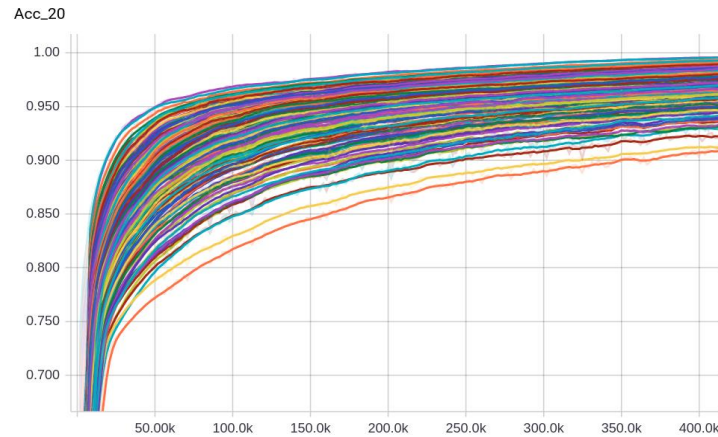
Training dataset

Validation dataset

Loss function



Accuracy<sub>20%</sub>

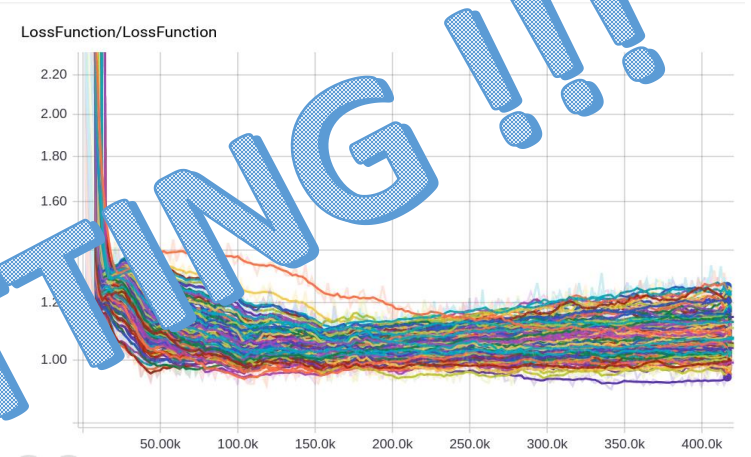
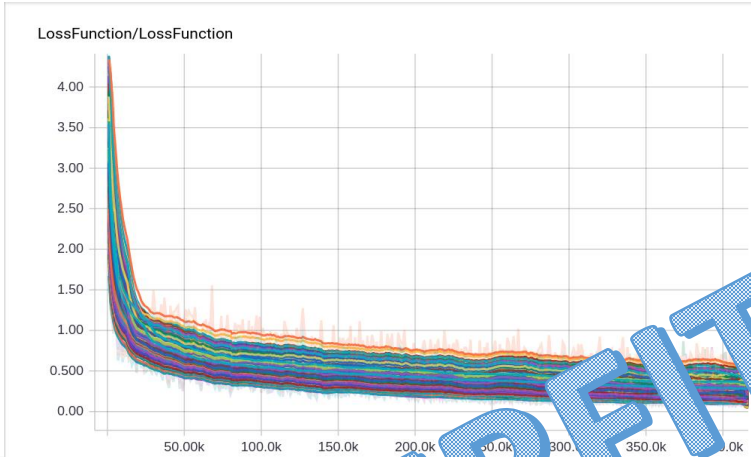


# Results: Overall Performance

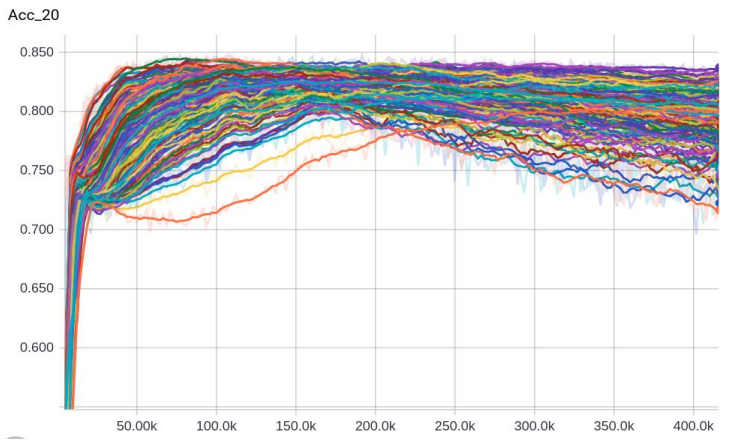
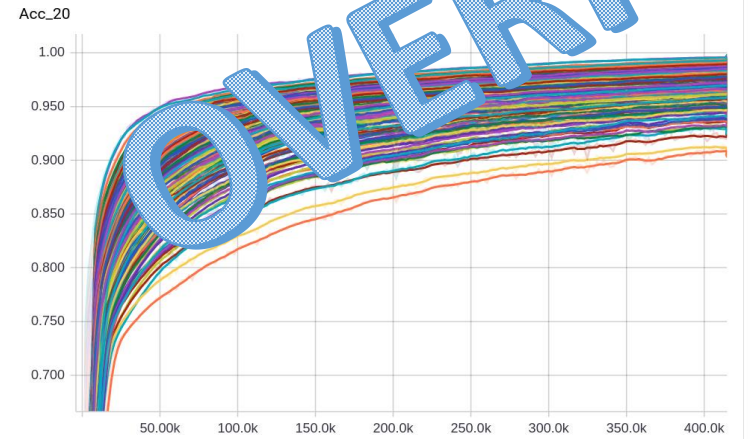
Training dataset

Validation dataset

Loss function



Accuracy<sub>20%</sub>



OVERFITTING!!!

# Results: Overfitting solutions

- ✓ Increase size of training dataset.
- ✓ Change of activation functions.
- ✓ User of Regularization techniques.
- ✓ Early Stopping.



# Results: Overfitting solutions

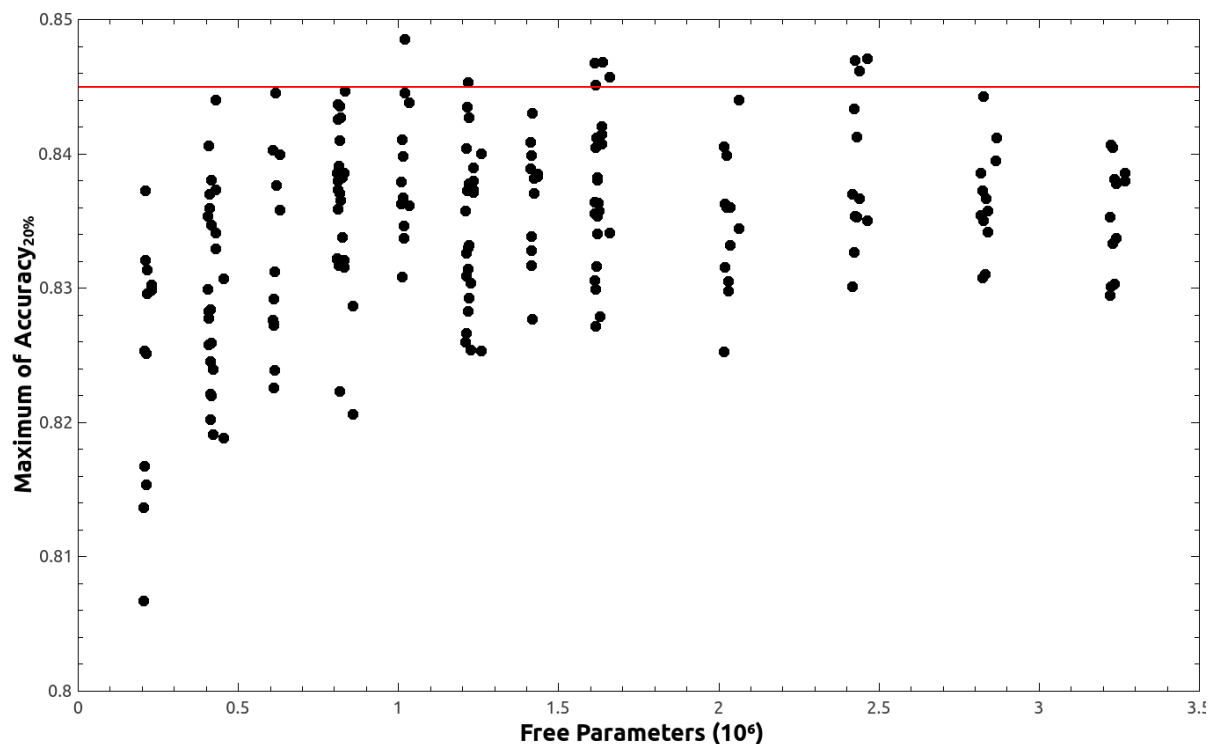
- ✓ Increase size of training dataset.
- ✓ Change of activation functions.
- ✓ User of Regularization techniques.
- ✓ Early Stopping.

**EARLY STOPPING !!!**



# Results: Early Stopping

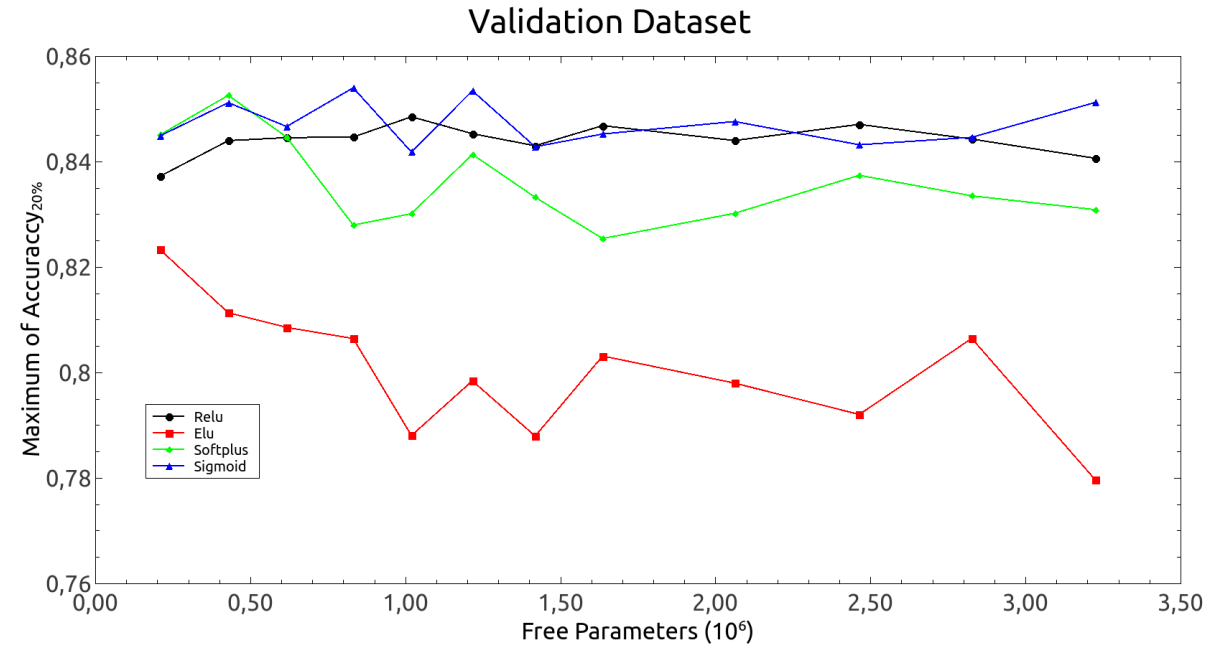
## Top 10 models



	$w_{c1}$	$n_{c1}$	$w_{c2}$	$n_{c2}$	$n_{fc1}$	Max Accuracy <sub>20%</sub>	Iteration Step	Free Parameters
<b>Model-108</b>	5	16	5	32	640	0.8485	108,264	1,018,690
<b>Model-189</b>	5	32	5	64	768	0.8471	68,706	2,462,850
<b>Model-061</b>	5	8	5	64	768	0.847	106,182	2,423,826
<b>Model-175</b>	5	32	5	32	1024	0.8469	68,706	1,635,170
<b>Model-039</b>	5	8	3	32	1024	0.8468	191,544	1,611,250
<b>Model-125</b>	5	16	5	64	768	0.8462	68,706	2,436,834
<b>Model-187</b>	5	32	5	64	512	0.8457	83,280	1,659,266
<b>Model-114</b>	5	16	3	64	384	0.8453	158,232	1,215,074
<b>Model-047</b>	5	8	5	32	1024	0.8452	77,034	1,615,346
<b>Model-171</b>	5	32	5	32	512	0.8447	104,100	830,818

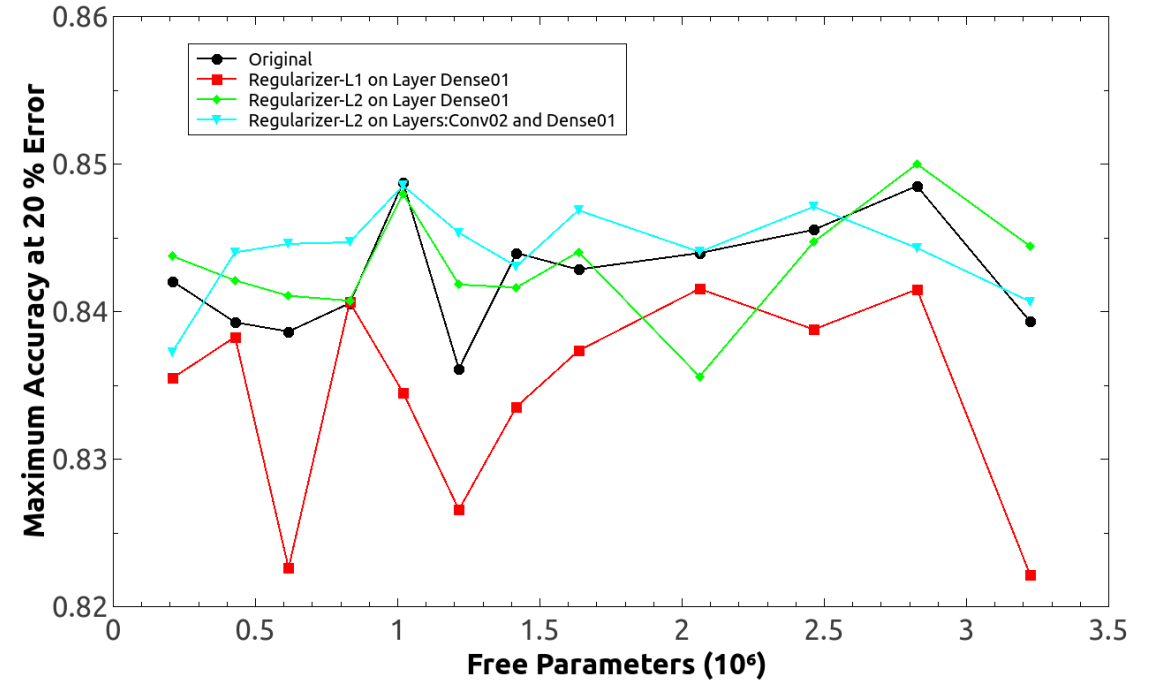
# Results: Activation Function

	wc1	nc1	wc2	nc2	nfc1	Free Parameters
Model-040	5	8	5	32	128	207730
Model-137	3	32	5	32	256	428130
Model-074	3	16	5	32	384	616258
Model-171	5	32	5	32	512	830818
Model-108	5	16	5	32	640	1018690
Model-114	5	16	3	64	384	1215074
Model-166	5	32	3	32	896	1417698
Model-175	5	32	5	32	1024	1635170
Model-188	5	32	5	64	640	2061058
Model-189	5	32	5	64	768	2462850
Model-062	5	8	5	64	896	2825618
Model-119	5	16	3	64	1024	3224034



# Results: Regularization

	wc1	nc1	wc2	nc2	nfc1	Free Parameters
Model-040	5	8	5	32	128	207730
Model-137	3	32	5	32	256	428130
Model-074	3	16	5	32	384	616258
Model-171	5	32	5	32	512	830818
Model-108	5	16	5	32	640	1018690
Model-114	5	16	3	64	384	1215074
Model-166	5	32	3	32	896	1417698
Model-175	5	32	5	32	1024	1635170
Model-188	5	32	5	64	640	2061058
Model-189	5	32	5	64	768	2462850
Model-062	5	8	5	64	896	2825618
Model-119	5	16	3	64	1024	3224034



$$L1 = \lambda \sum_{i=1}^k |w_i|$$

$$\lambda(L1) = 0.001$$

$$L2 = \lambda \sum_{i=1}^k w_i^2$$

$$\lambda(L2) = 0.01$$

# Summary and Conclusions

An hyper-parametric search (# filters, filter size and #neurons) was conducted to obtain best model for monitoring LMD process based on MWIR images.

Thanks to Finis Terrae II HPC CESGA infrastructure the time-to –solution was reduced from 40 days to 70 hours.

Time factor decrease of 14 but could increase to more that 100!!.

Additional approximations like changing activation functions or using regularizers can improve model performance.



FORTISSIMO



THANKS FOR  
YOUR ATTENTION !!!

Co-funded by  
the European Union



This project has received funding from the European Union's H2020  
research and innovation programme under grant agreement No 680481