

A decorative graphic on the left side of the slide, consisting of a network of blue and grey lines and circles, resembling a circuit board or a neural network diagram.

DEEP LEARNING FOR BACTERIA IDENTIFICATION USING RAMAN SPECTROSCOPY

ADVANCED MACHINE LEARNING

FEBRUARY 2021

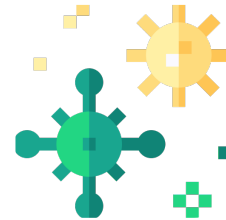
Davide Marchetti / 815990 / d.marchetti8@campus.unimib.it

Francesco Stranieri / 816551 / f.stranieri1@campus.unimib.it

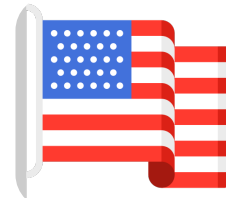
INTRODUCTION

- *Raman spectroscopy* has the potential to identify the species of bacteria and its antibiotic susceptibility
- Classical ML techniques have been widely applied, while relatively little work was done in adapting DL models

Bacterial infections



6.7 million deaths every year



\$33 billion for annual healthcare spending in the USA

PROJECT GOALS

30-class task

Develop an efficient CNN architecture

Compare it with a state-of-the-art CNN

15-class task

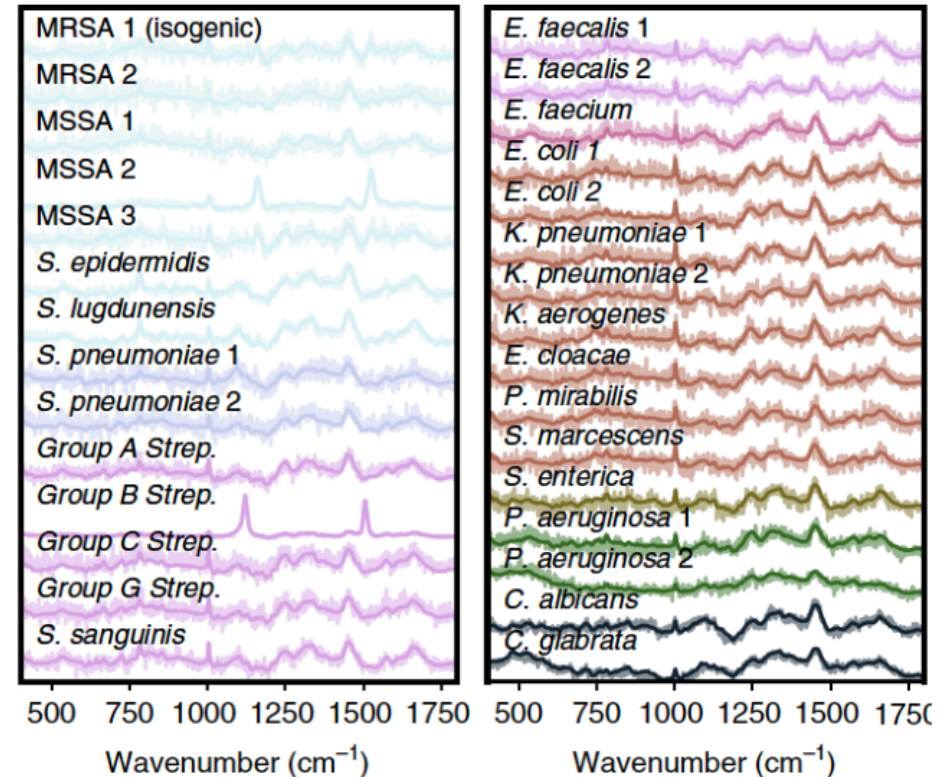
Fine-tuning process on our CNN model

Compare it against two ML algorithms

DATASET

- Training dataset of 60 000 spectra (2 000 per class)
- Fine-tuning and (independent) test datasets each of 3 000 spectra (100 per class)
- 30 bacterial isolate classes

94% of all bacterial infections treated at Stanford Hospital in the years 2016–2017



30-CLASS TASK – CHALLENGE

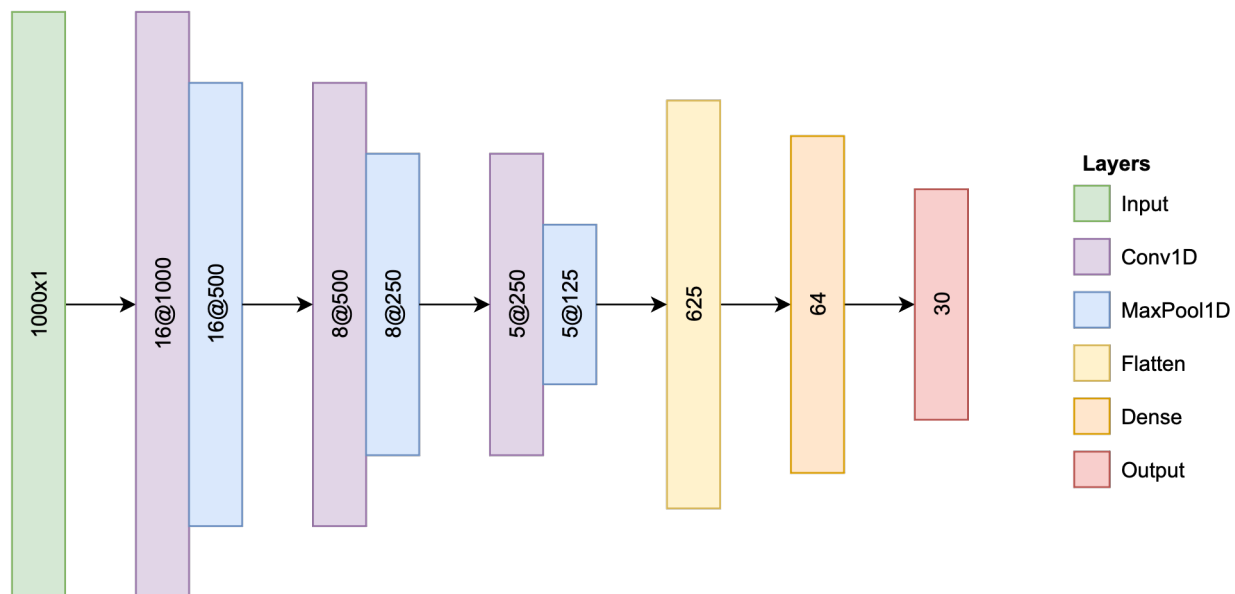
- Is the reference CNN architecture *overkill*?

Reference CNN

- 26 layers
- 1 340 000 parameters

30-CLASS TASK – CNN

- Is the reference CNN architecture *overkill*?



Reference CNN

- 26 layers
- 1 340 000 parameters

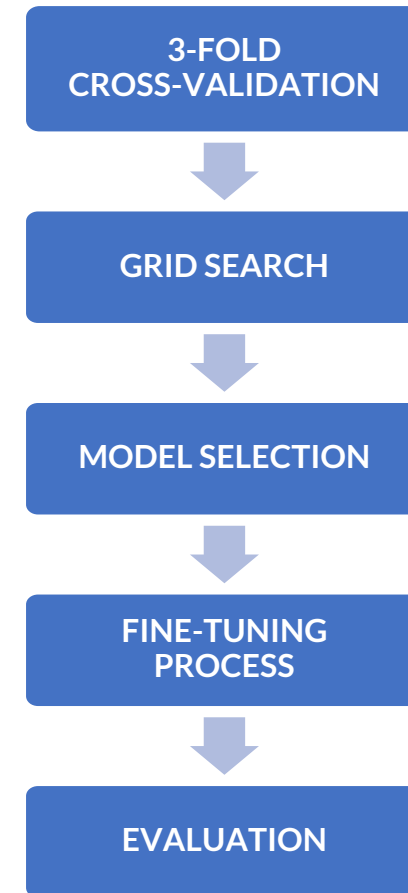
Proposed CNN

- 6 layers
- 42 000 parameters (96.8% less)

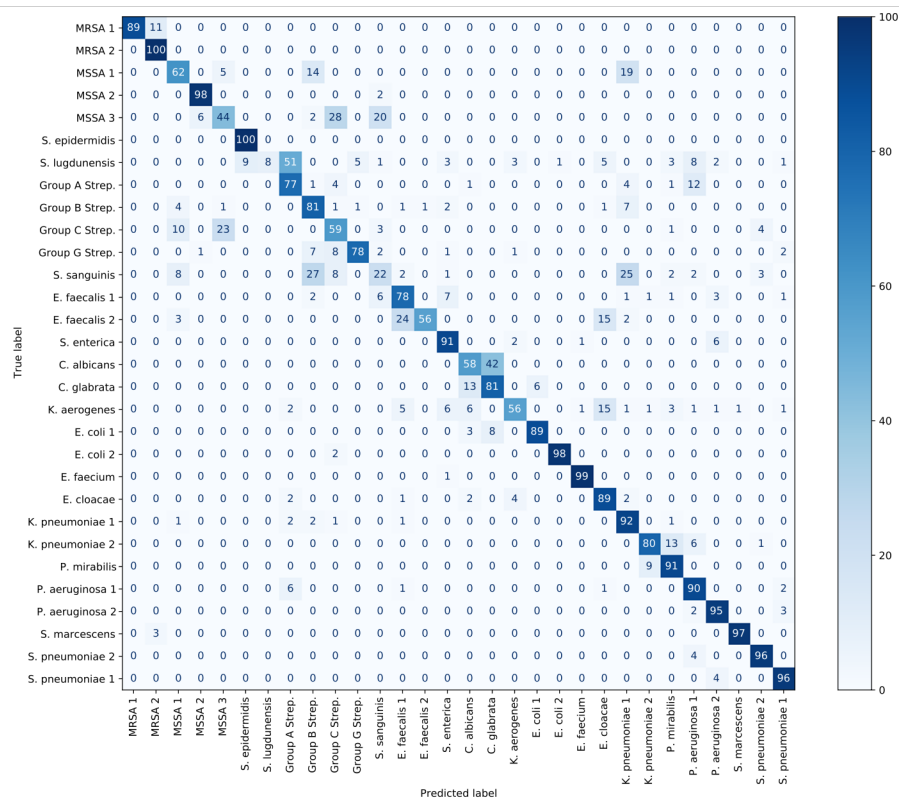
30-CLASS TASK – EXPERIMENTAL PLAN

- *3-fold cross-validation* on the training data via *grid search* (144 experiments in total)
- Average test accuracy to perform *model selection*
- Split *fine-tuning* data into 80% train and 20% validation to further train and detect overfitting
- *Evaluation* on test data

Hyperparameter	Values
batch_size	16, 32, 64
conv_layer	2, 3
filters	16, 32
kernel_size	3, 5
units	256, 512, 1024
dropout_rate	.3, .5

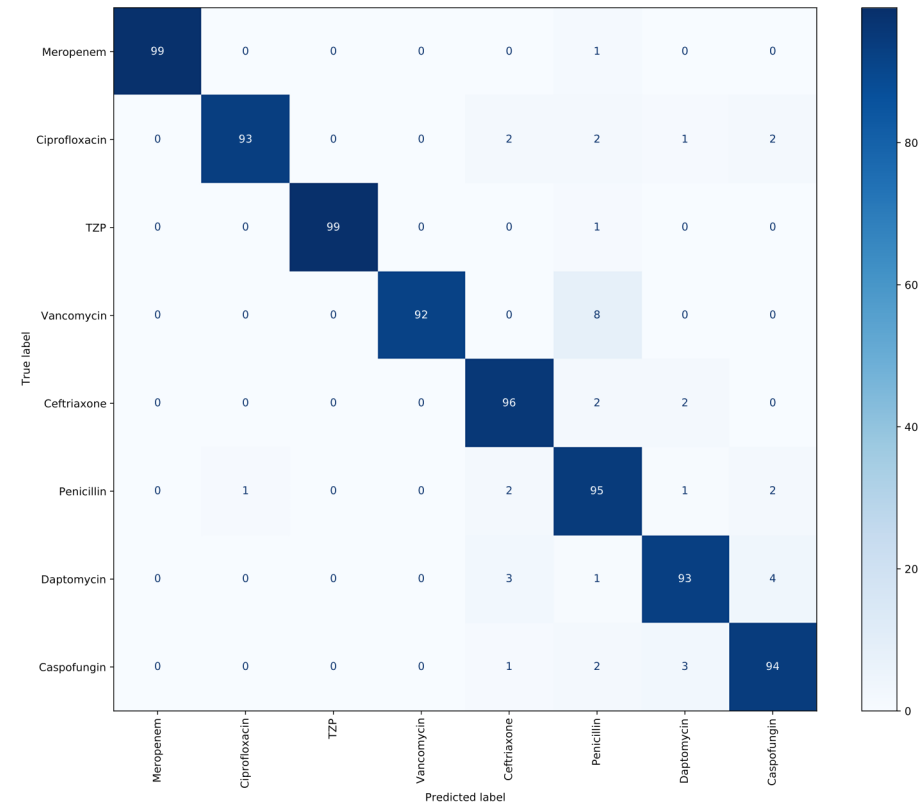


30-CLASS TASK – RESULTS



REFERENCE:
82.2±0.3%

PROPOSED:
78.3%


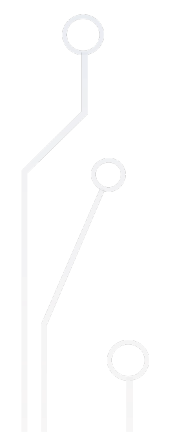


REFERENCE:
97.0±0.3%

PROPOSED:
94.9%



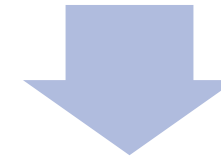
15-CLASS TASK – CHALLENGE

- Can SVM and MLP be used on a smaller 15-class classification task?
 - Are the *performances comparable* to those achieved by our CNN model?
- 
- 

15-CLASS TASK – CNN

- Can SVM and MLP be used on a smaller 15-class classification task?
- Are the *performances comparable* to those achieved by our CNN model?
- For the CNN model we exploit the *fine-tuning process* on the 15-class

FINE-TUNING
PROCESS

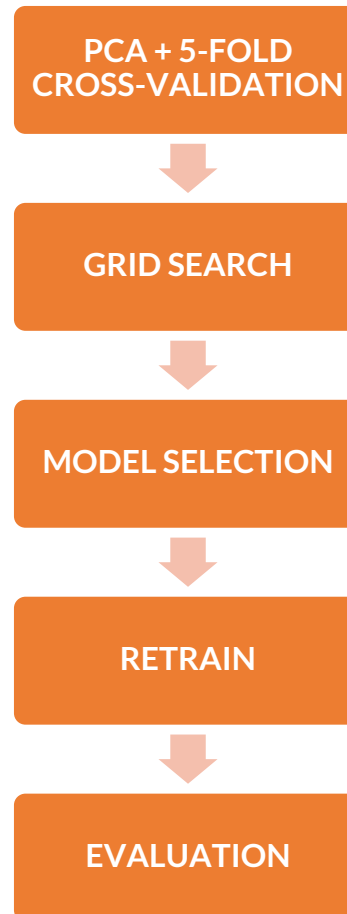


EVALUATION

15-CLASS TASK – SVM

- *PCA* to reduce input dimension from 1000 to 20
- *5-fold cross-validation* on fine-tuning dataset via *grid search* (32 experiments in total)
- Average test accuracy to perform *model selection*
- *Retraining* on the entire fine-tuning dataset
- *Evaluation* on test data

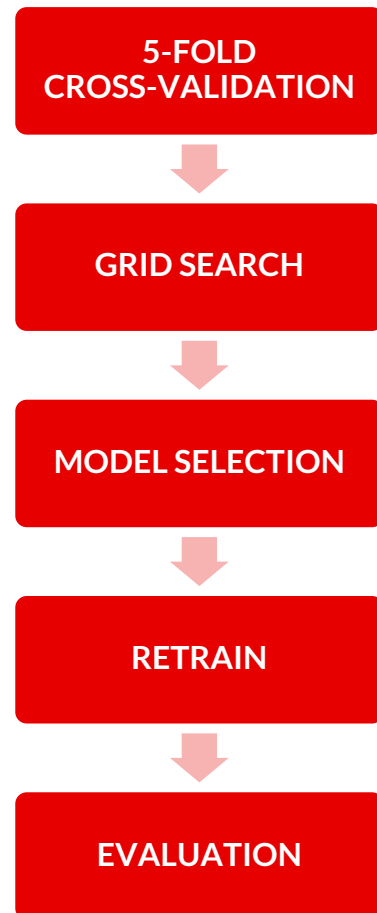
Hyperparameter	Values
kernel	rbf, <i>linear</i>
gamma	0, .001, .0001
C	1, 10, 100, 1000



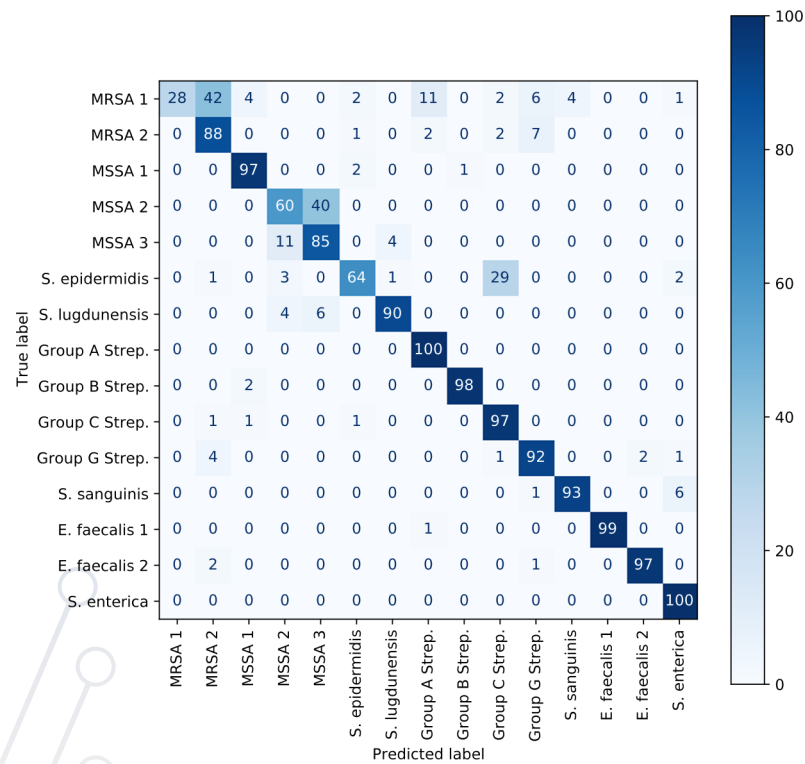
15-CLASS TASK – MLP

- *5-fold cross-validation* on fine-tuning dataset via *grid search* (36 experiments in total)
- Average test accuracy to perform *model selection*
- Split fine-tuning data into 80% train and 20% validation to *retrain* the selected model and detect overfitting
- *Evaluation* on test data

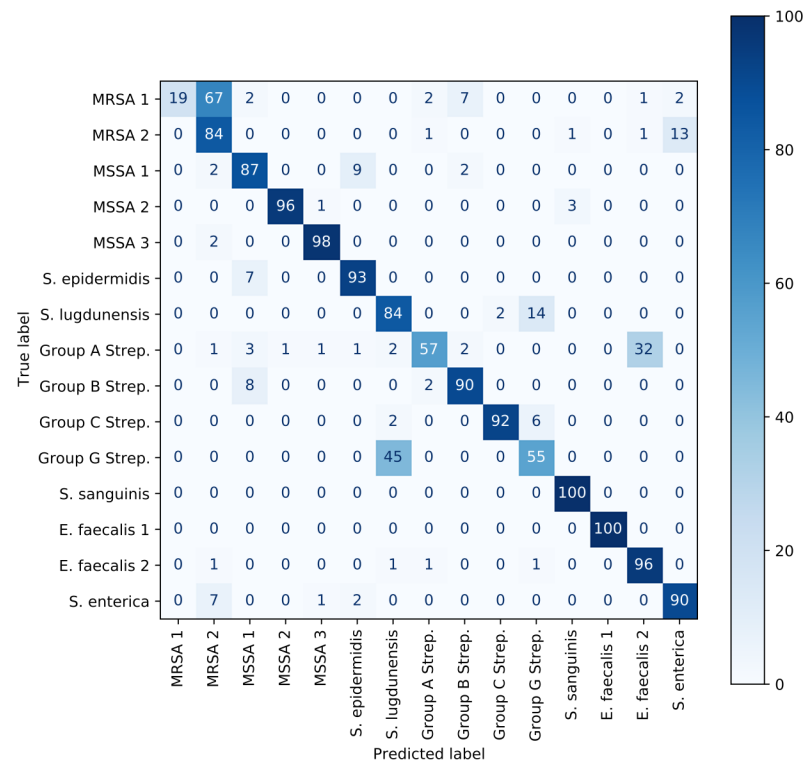
Hyperparameter	Values
batch_size	16, 32, 64
units	256, 512
hidden_layers	1, 2, 3
dropout_rate	.3, .5



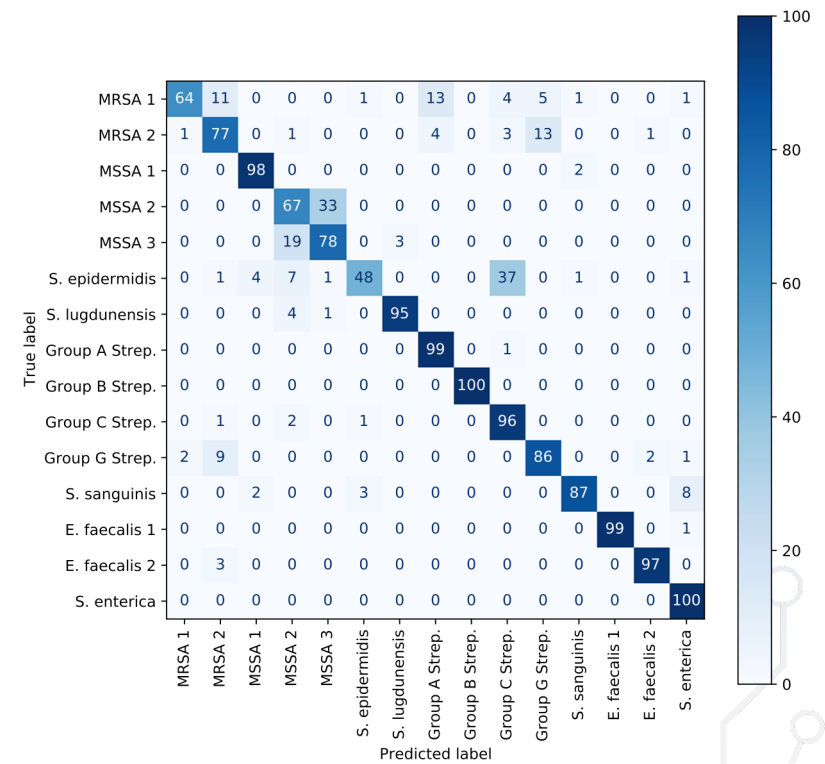
15-CLASS TASK – BACTERIAL RESULTS



CNN:
85.1%

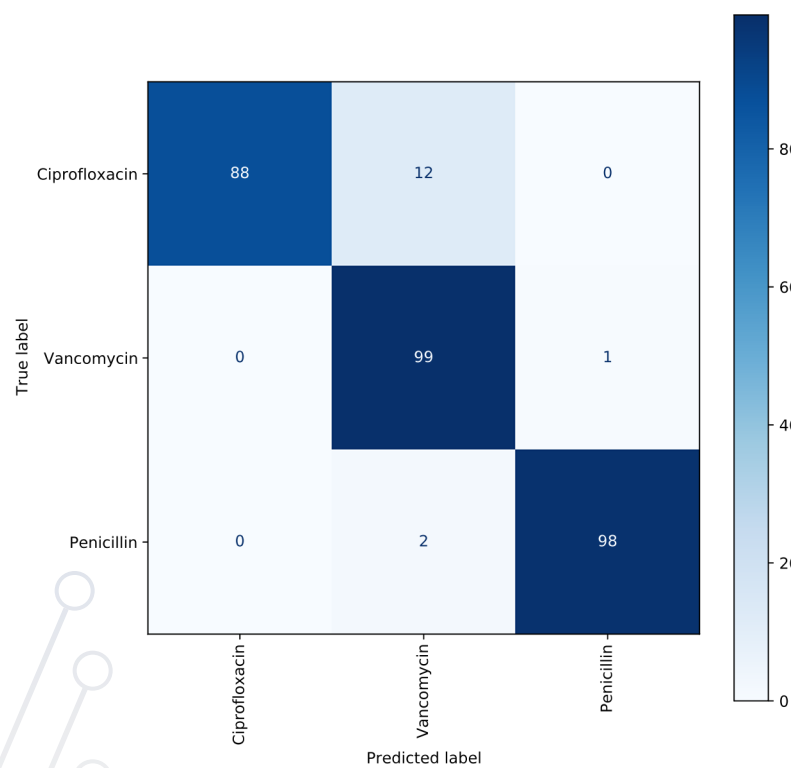


SVM:
82.7%

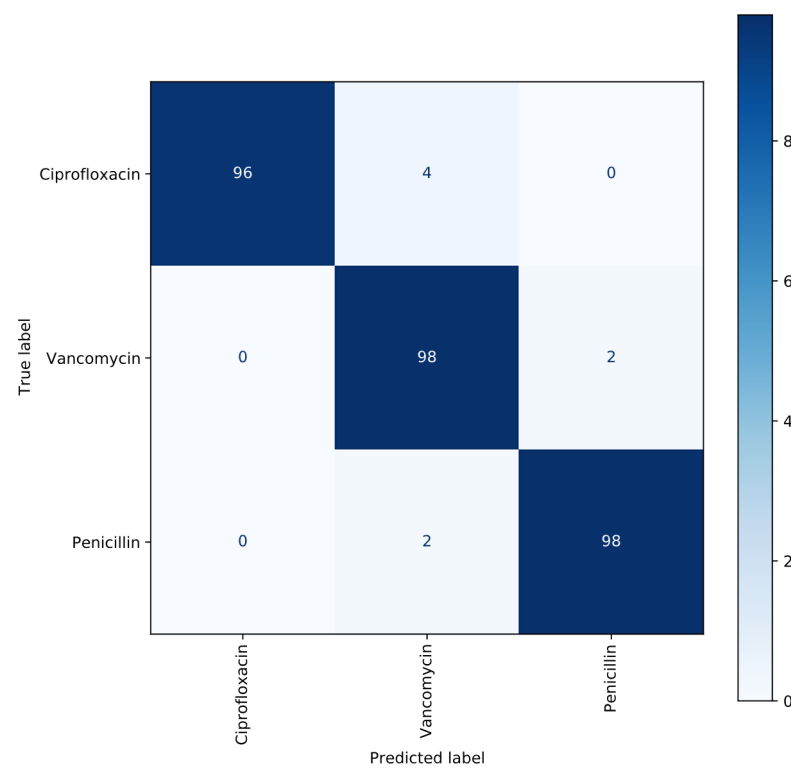


MLP:
85.0%

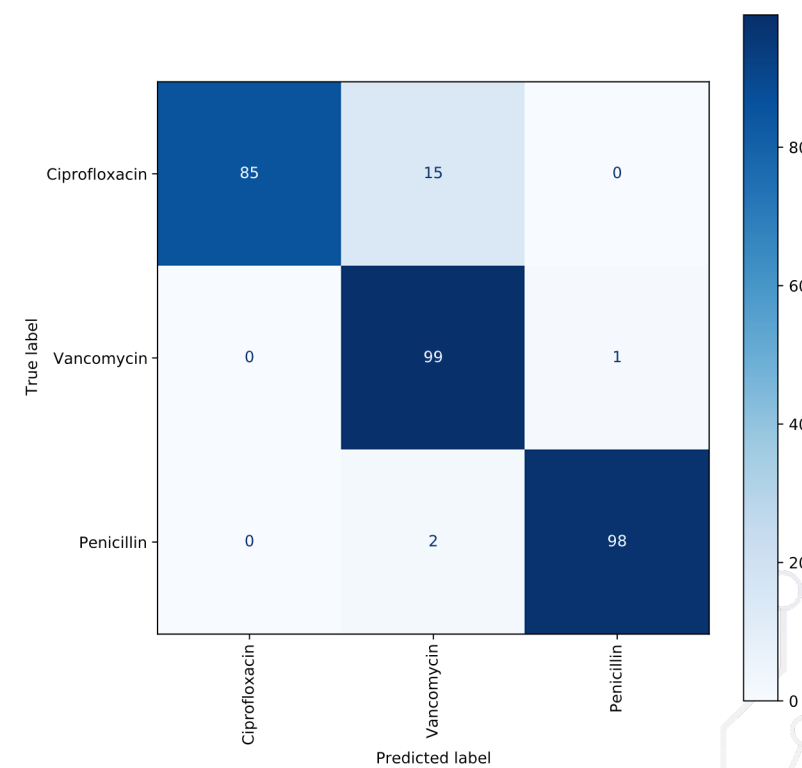
15-CLASS TASK – ANTIBIOTICS RESULTS



CNN:
98%



SVM:
97.7%



MLP:
97.5%

CONCLUSIONS

- On the 30-class task, the presented CNN architecture achieves *comparable performance* with a state-of-the-art CNN, although with *less parameters*
- On the 15-class task, classical ML techniques reported *good results*
- In general, *misclassifications* are mostly within antibiotic groupings, and thus do not affect the treatment outcome

DL techniques applied to Raman spectra would allow for accurate and targeted treatment of bacterial infections within hours, reducing healthcare costs, antibiotics misuse and improving patient outcomes

REFERENCES

- F. Lussier, V. Thibault, B. Charron, G. Q. Wallace, and J.-F. Masson, "Deep learning and artificial intelligence methods for raman and surface-enhanced raman scattering," *TrAC Trends in Analytical Chemistry*, vol. 124, p. 115796, 2020.
- C. Fleischmann, A. Scherag, N. K. Adhikari, C. S. Hartog, T. Tsaganos, P. Schlattmann, D. C. Angus, and K. Reinhart, "Assessment of global incidence and mortality of hospital-treated sepsis. current estimates and limitations," *American journal of respiratory and critical care medicine*, vol. 193, no. 3, pp. 259–272, 2016.
- R. DeAntonio, J.-P. Yarzabal, J. P. Cruz, J. E. Schmidt, and J. Kleijnen, "Epidemiology of community-acquired pneumonia and implications for vaccination of children living in developing and newly industrialized countries: A systematic literature review," *Human vaccines & immunotherapeutics*, vol. 12, no. 9, pp. 2422–2440, 2016.
- C. M. Torio and B. J. Moore, "National inpatient hospital costs: the most expensive conditions by payer, 2013: statistical brief# 204," 2016.
- R. P. Dellinger, M. M. Levy, A. Rhodes, D. Annane, H. Gerlach, S. M. Opal, J. E. Sevransky, C. L. Sprung, I. S. Douglas, R. Jaeschke et al., "Surviving sepsis campaign: international guidelines for management of severe sepsis and septic shock, 2012," *Intensive care medicine*, vol. 39, no. 2, pp. 165–228, 2013.
- A. Chaudhuri, P. Martin, P. Kennedy, R. Andrew Seaton, P. Portegies, M. Bojar, I. Steiner, and E. T. Force, "Efn guideline on the management of community-acquired bacterial meningitis: report of an efns task force on acute bacterial meningitis in older children and adults," *European journal of neurology*, vol. 15, no. 7, pp. 649–659, 2008.
- C.-S. Ho, N. Jean, C. A. Hogan, L. Blackmon, S. S. Jeffrey, M. Holodniy, N. Ba-naei, A. A. Saleh, S. Ermon, and J. Dionne, "Rapid identification of pathogenic bacteria using raman spectroscopy and deep learning," *Nature communications*, vol. 10, no. 1, pp. 1–8, 2019.
- J.-L. Vincent, J. Rello, J. Marshall, E. Silva, A. Anzueto, C. D. Martin, R. Moreno, J. Lipman, C. Gomersall, Y. Sakr et al., "International study of the prevalence and outcomes of infection in intensive care units," *Jama*, vol. 302, no. 21, pp. 2323–2329, 2009.