

# Your Title

Author 1, Author 2  
Chair of Statistics, University of Göttingen

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## **Abstract**

The text of your abstract. 200 or fewer words.

*Keywords:* 3 to 6 keywords

# 1 Introduction

Describe the dataset and define the problem. Do not explain in detail the basics of ANNs as we did in the introductory lecture.

(Goodfellow et al., 2016)

Hope et al. (2017); Chollet (2018)



Figure 1: Caption

## 2 Methods

Describe the architecture of your net, the training and validation process, regularisation methods you used, ...

Beispiel des von uns gewünschten Pseudo-Code Layouts:

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**Algorithm 1** Major Iteration: Bound-Constraint Lagrangian

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**Require:** initial weights  $\mathbf{w}_0 \in \mathbb{R}^K$ , multipliers  $\lambda^0$ , initial value penalty parameter  $\gamma_0 > 0$  and given total convergence tolerances for  $\eta_*$  and  $\omega_*$

**for**  $k = 1, 2, \dots$  **do**

    Using SQP finding an approximate solution for  $w_0$  of the subproblem

$$\|\mathbf{w}_k - P(\mathbf{w}_k - \nabla_{\mathbf{w}} \mathcal{L}_A(\mathbf{w}_k, \lambda^k; \gamma_k), l, u)\| \leq \omega_k$$

**if**  $\|c(\mathbf{w}_k)\| \leq \eta_k$  **then**

    testing for convergence of given subproblem

**if**  $\|c(\mathbf{w}_*)\| \leq \eta_* \wedge q_k(\mathbf{w}) \leq \omega_*$  **then**

**stop** approximate solution of  $w_k$  found

**end if** solution holds

    Update the underlying multipliers

$$\lambda^{k+1} = \lambda^k - \gamma_k c(x_k)$$

$$\gamma_{k+1} = \gamma_k$$

**else**

    If no valid solution was found and increase the penalties

$$\lambda^{k+1} = \lambda^k$$

$$\gamma_{k+1} = 100\gamma_k$$

**end if**

**end for**

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## 3 Results

Describe the results of your analysis.

Beispiel des gewünschten Tabellen Layouts:

Table 1: Calculated weight for M1 ( $w_1$ ) for different given variance ratios

$\tau_1^2$	Mean	Std.Error
0	0.069	0.059
0.25	0.205	0.096
0.50	0.499	0.125
0.75	0.795	0.097
1	0.933	0.061

## 4 Conclusion

Discuss your results.

# A Appendix

## A.1 ..

## References

- Chollet, F. (2018). *Deep learning with Python*. Manning Publications Co, Shelter Island, New York. OCLC: ocn982650571.
- Goodfellow, I., Bengio, Y., and Courville, A. (2016). *Deep Learning*. MIT Press.
- Hope, T., Resheff, Y. S., and Lieder, I. (2017). Learning TensorFlow.