

# TEMPLATE FOR A PROJECT REPORT

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Report for a project within the course CS-E4740 - Federated Learning.

## ABSTRACT

**Instructions (Remove before submission).** Provide a concise summary of the project, including the FL application, how you modelled it as a FL network, the GTV minimization on top of it and the FL algorithms used.

**Index Terms**— Federated Learning, Networks, Personalized Machine Learning, Trustworthy AI

## 1. INTRODUCTION

**Instructions (Remove before submission)** This section introduces the background and motivation of your FL project. It should include:

- A real-life scenario motivating your FL application, consisting of networked devices that train personalized models (see Lecture “FL Networks”).
- A summary of state-of-the-art methods relevant to your project.
- A brief outline of the structure of your report.

## 2. PROBLEM FORMULATION

**Instructions (Remove before submission).** A core task for the project is the modelling of your FL application as an federated learning (FL) network (see Lecture “FL Networks” and [2, Ch. 3]). This section must clearly state and explain:

- **Nodes:** How many and what real-world devices (smartphones, smart light-bulbs, smart meters, a fitness tracker or a desktop computer) do they represent.
- **Local Models:** Describe the models used at each node. You can also state several different candidate models for a node.
- **Loss Functions:** For each node, specify the local loss functions for training the corresponding local model.
- **Edges:** How do you choose the edges and their weights? (see [2, Sec. 7])

## 3. METHODS

**Instructions (Remove before submission).** Design FL algorithms for your application by formulating FL as GTVMin which is solved using distributed optimization methods. Clearly state and explain:

- Your choice for the variation measure. For example, when using parametric local models you can use  $\phi(\mathbf{w}^{(i)} - \mathbf{w}^{(i')})$  to measure the variation across edge  $\{i, i'\}$  (see Lecture “FL Design Principle” and [2, Ch. 3]).
- Your choice for FL algorithm, which must be implemented as message passing across the edges of the FL network (see Lecture “FL Algorithms” and [2, Ch. 5]).

## 4. NUMERICAL EXPERIMENTS

**Instructions (Remove before submission).** Provide details on the implementation of the FL algorithms and analyze the results. Clearly state and explain

- the used data sources. One example for a data source is the open data interface of the Finnish Meteorological institute [3].
- Your approach to model validation, selection and diagnosis (see [1, Sec. 6.6]). In particular,
  - how did you implement model validation?
  - If you have considered different options for the local model used by some node  $i$ , how did you choose between those options?
  - how happy are you with the performance of the resulting trained local models? can you think of a baseline to compare against?
- for each trained local model, state and compare the average loss on the training set and the validation set.
- how you obtained the final trained local models. This includes

- the choice of FL network (including choice for local model and edge set)
  - the choice for the variation measure (penalty function for parametric models).
  - the FL algorithm (which must be implemented as message passing)
- For the finally obtained trained local models, report the average loss on a test set (which has neither been used for training or validation of local models).

**Important:** Your report must be accompanied by a Python notebook that allows reproduce the numerical results. Try to use only basic libraries such that the notebook can be executed without hassle on <https://jupyter.cs.aalto.fi/>.

## 5. CONCLUSION

Instructions (Remove before submission):

- Discuss whether the obtained numerical results suggest that the problem is solved satisfactorily or if there is room for improvement.
- Identify limitations of the methods used and suggest potential improvements.

## 6. REFERENCES

- [1] A. Jung, *Machine Learning: The Basics*, Springer Singapore, 1st edition, Feb. 2022. working draft: <https://github.com/alexjungaalto/MachineLearningTheBasics/blob/master/MLBasicsBook.pdf>
- [2] A. Jung, “Federated Learning: From Theory to Practice,” Aalto, 2025. <https://github.com/alexjungaalto/FederatedLearning/blob/main/material/FLBook.pdf>
- [3] Finnish Meteorological Institute, “Open Data,” 2025, <https://en.ilmatieteenlaitos.fi/open-data>, Accessed: 2025-02-27.