

CS-E4740 - Federated Learning

L0 - Course Logistics

Assoc. Prof. Alexander Jung

Spring 2026

Calendar



Glossary



Book



GitHub



At a Glance

- ▶ 5 credits (approx. 130 hours of work)
- ▶ **fully online - no mandatory attendance**
- ▶ Part I (Feb.,Mar.): 8 lectures and assignments
- ▶ Part II (Apr.,May): student project with peer review

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Prerequisites

- ▶ **Linear Algebra.** Vectors $\mathbf{w} \in \mathbb{R}^d$, matrices $\mathbf{Q} \in \mathbb{R}^{d \times d}$, norms $\|\mathbf{w}\|_2$.
- ▶ **Multivariable Calculus.** Smooth functions $f(\mathbf{w})$ and their gradients $\nabla f(\mathbf{w})$.
- ▶ **Basic machine learning (ML).** Empirical risk minimization (ERM)
$$\min_{h \in \mathcal{H}} (1/m) \sum_{r=1}^m L((\mathbf{x}^{(r)}, y^{(r)}), h).$$
- ▶ **Python.** Basic coding skills and familiarity with libraries `numpy` and `scikit-learn`

see relevant terms and references in our [Aalto Dictionary of ML](#)

Learning Goals

After completing this course, you can

- ▶ represent federated learning (FL) applications using graphs
- ▶ formulate FL as optimization over graphs
- ▶ design FL algorithms via distributed optimization
- ▶ build trustworthy federated learning systems (FL systems)

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Related Courses - Bare Necessities

- ▶ **MS-A0001 - Matrix Algebra.** Introduction to linear algebra in \mathbb{R}^d . We will use \mathbb{R}^d as the main mathematical structure to reason about FL systems.
- ▶ **CS-C3240 - ML.** Teaches basic techniques for training a single ML model on a given dataset. FL extends this centralized setting to distributed inter-connected devices, each having access to its own dataset and model.
- ▶ **Data Analysis with Python.** Teaches how to implement basic ML methods in Python. Our course assignments require to implement (parts of) FL algorithms in Python.

Related Courses - Nice to Have

- ▶ **MS-C2105 - Introduction to Optimization.** Teaches basic concepts for the design and analysis of optimization methods. Our course formulates FL as an optimization problem. FL algorithms are obtained, in turn, by applying optimization methods to solve this problem.
- ▶ **ELEC-E5424 - Convex optimization.** Teaches advanced tools such as convergence analysis of gradient-based methods for the study and design of FL algorithms.

Related Courses - Follow Up

- ▶ **ELEC-E7120 - Wireless Systems.** Discusses the fundamentals of radio communications which can be used to implement FL algorithms.
- ▶ **ELEC-E8102 - Distributed and Intelligent Automation Systems.** Discusses automation systems consisting of interconnected sensors and actuators. We can use FL to train predictive models used by these devices.

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- ▶ Part I (Feb.,Mar.): 8 lectures and assignments
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Lecture Schedule

Lecture	Day	Topic
L1	Mon 23.02	From ML to FL
L2	Wed 25.02	FL Design Principle
L3	Mon 02.03	FL Algorithms I
L4	Wed 04.03	FL Algorithms II
L5	Mon 09.03	Federated Clustering
L6	Wed 11.03	Explainable FL
L7	Mon 16.03	Cyber-Security in FL
L8	Wed 18.03	Privacy in FL

Assignment Workflow

- ▶ one assignment per lecture
- ▶ small Python tasks and conceptual questions
- ▶ released after the lecture
- ▶ assessment via multiple-choice questions
- ▶ there is only one single attempt for each quiz

JupyterHub installation

you can use your own Python installation

instead you could also use <https://jupyter.cs.aalto.fi>:

- ▶ log in with Aalto credentials
- ▶ select "Python: General use (JupyterLab)"
- ▶ you now have a virtual machine in your browser

Assignment Quizzes

The screenshot shows the course page for CS-E4740 - Federated Learning D, Lecture, 23.2.2026-27.5.2026. The left sidebar contains a navigation menu with 'Main course page', 'General' (Announcements, Course Discussion Forum), 'Assignments' (From ML to FL, FL Design Principle), and 'Course Project'. The main content area is titled 'Assignments' and features a yellow icon for 'From ML to FL'. Below the icon, it states: 'This quiz tests your solutions to the assignment for the lecture "From ML to FL".'

- ▶ no minimum point requirement for any individual quiz
- ▶ for each quiz, you have **one single attempt**
- ▶ you can book a review meeting at end of the course
- ▶ the review meeting can also be used to rectify grading

Assignment Schedule

	Released	Due	Topic
A1	Mon 23.02	Sun 08.03	From ML to FL
A2	Wed 25.02	Sun 08.03	FL Design Principle
A3	Mon 02.03	Sun 15.03	FL Algorithms I
A4	Wed 04.03	Sun 15.03	FL Algorithms II
A5	Mon 09.03	Sun 22.03	Federated Clustering
A6	Wed 11.03	Sun 22.03	Explainable FL
A7	Mon 16.03	Sun 29.03	Cyber-Security in FL
A8	Wed 18.03	Sun 29.03	Privacy in FL

on average, you have to complete **one assignment per week**

Project Schedule (April–May)

Milestone	Date
Project kickoff event	Wed 01.04.2026
First submission DL	Sun 26.04.2026
Peer Review – Start	Mon 27.04.2026
Peer Review – End	Sun 03.05.2026
Second submission DL	Sun 17.05.2026
Peer Review – Start	Mon 18.05.2026
Peer Review – End	Wed 27.05.2026

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Graded Activities

- ▶ **8 Assignments.** One for each lecture. Completion of the assignments is tested via multiple choice quizzes.
- ▶ **Student Project.** Apply FL techniques to weather forecasting. Two submission rounds, each involving peer grading.

Points Breakdown

Component	Points
Assignments (one for each lecture)	40
Student Project (60 points)	
First Submission	5
Peer Review 1	4
Second submission (incl. presentation)	43
Peer Review 2	8
Total	100

- ▶ no minimum requirement for any individual task
- ▶ you have the right to a review meeting at end of course

From Points to Grade

- ▶ **grade 1** for 50-59 points
- ▶ **grade 2** for 60-69 points
- ▶ **grade 3** for 70-79 points
- ▶ **grade 4** for 80-89 points
- ▶ **top grade 5** for at least 90 points

Project Workflow

Step	What you do
1st submission	submit first version of project report.
Peer Review 1	review assigned drafts and receive feedback
2nd submission	submit revised report +response letter+ recorded slide talk
Peer Review 2	review improved reports

Project - First Submission

- ▶ federated learning network (FL network) representing stations of Finnish Meteorological Institute (FMI)
- ▶ you can freely define data points but their features and labels must be derived from raw FMI data

Your task:

- ▶ design and evaluate **at least one FL method**
- ▶ quantitative evaluation and discussion

Deliverables and assessment:

- ▶ IEEE-style report (template provided) + reproducible code
- ▶ anonymous, rubric-based peer review (TA moderated)

Project – Second Submission

Goal: Improve your work based on peer feedback from the first submission.

Deliverables:

- ▶ Revised IEEE-style report (pdf)
- ▶ Response-to-reviews letter (pdf)
- ▶ Project presentation (mp4)

Response letter requirements:

- ▶ for each reviewer comment, explain: (i) what you changed, (ii) where in the report, (iii) why
- ▶ if you disagree with comment: explain briefly

Assessment:

- ▶ Second round of anonymous peer review (TA moderated)

Project – Recorded Slide Presentation

Recorded slide talk (voice-over) that summarizes your project

Specifications:

- ▶ submitted as mp4 via MyCourses
- ▶ max. duration: 5:00 minutes
- ▶ max. file size: 200 MB

Project – Recorded Slide Presentation (ctd.)

Expected content (suggested structure):

- ▶ Problem Formulation (30–45s)
- ▶ FL algorithm (60–90s)
- ▶ Experimental setup (45–60s)
- ▶ Results (90–120s)
- ▶ Limitations and Future Work (20–40s)

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This course follows the **Aalto University Code of Conduct** ([see here](#)).

We want to ensure **a safe learning environment** for everyone.

There are two simple principles guiding all course activities.

Rule 1 — Be Honest

- ▶ This course involves a significant amount of **independent work**:
 - ▶ weekly assignments
 - ▶ student project (report + recorded presentation)
 - ▶ peer review of other students' work
- ▶ **Your submissions must reflect your own understanding.**
- ▶ Plagiarism undermines fair assessment.
- ▶ We randomly invite students to Zoom meetings where they **need to explain their submitted work.**

Rule II — Be Respectful

This course aims to provide a **safe, inclusive, and respectful space**.

Respect applies to all course-related interactions, e.g., during lectures, peer reviews, discussion forum.

Disrespectful, discriminatory, or harassing behaviour has no place in this course and will be addressed according to university procedures.

ML Meetups

Subscribe · Add to calendar

