

# CS-E4740 Federated Learning

## "FL Project"

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# FL Project

pick **any application you like** and model it as a FL problem

you must:

- define/choose local datasets (train/val/test sets)
- define/choose local models
- define/choose loss functions
- define/choose edges and their weights
- define/choose discrepancy measure

# Local Datasets

 each local dataset  $\mathcal{D}^{(i)}$

consists of a

```
G.nodes[node_i]["ytrain"]  
G.nodes[node_i]["Xtrain"]
```

- train set: used to define local loss in GTVMin
- val set: used to select local models and/or edges
- test set: used for final performance assessment

# Local Models

 local dataset  $\mathcal{D}^{(i)}$

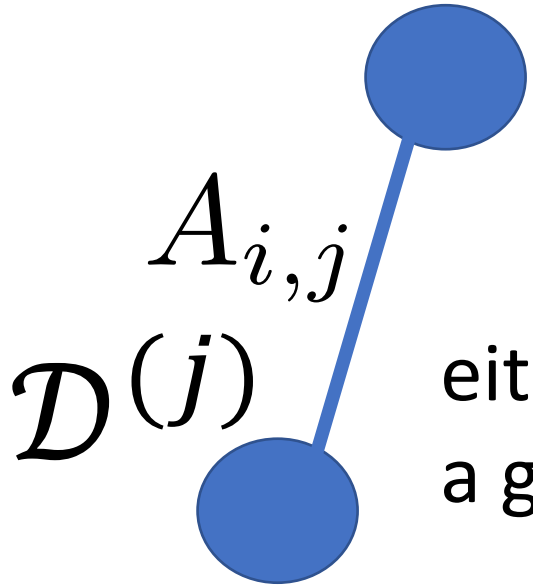
local model can be anything listed in MyCourses  
Section “FL Project”

```
Gin.nodes[node_i]["model"] = DecisionTreeRegressor(max_depth=4)
```

where did we choose loss function here?

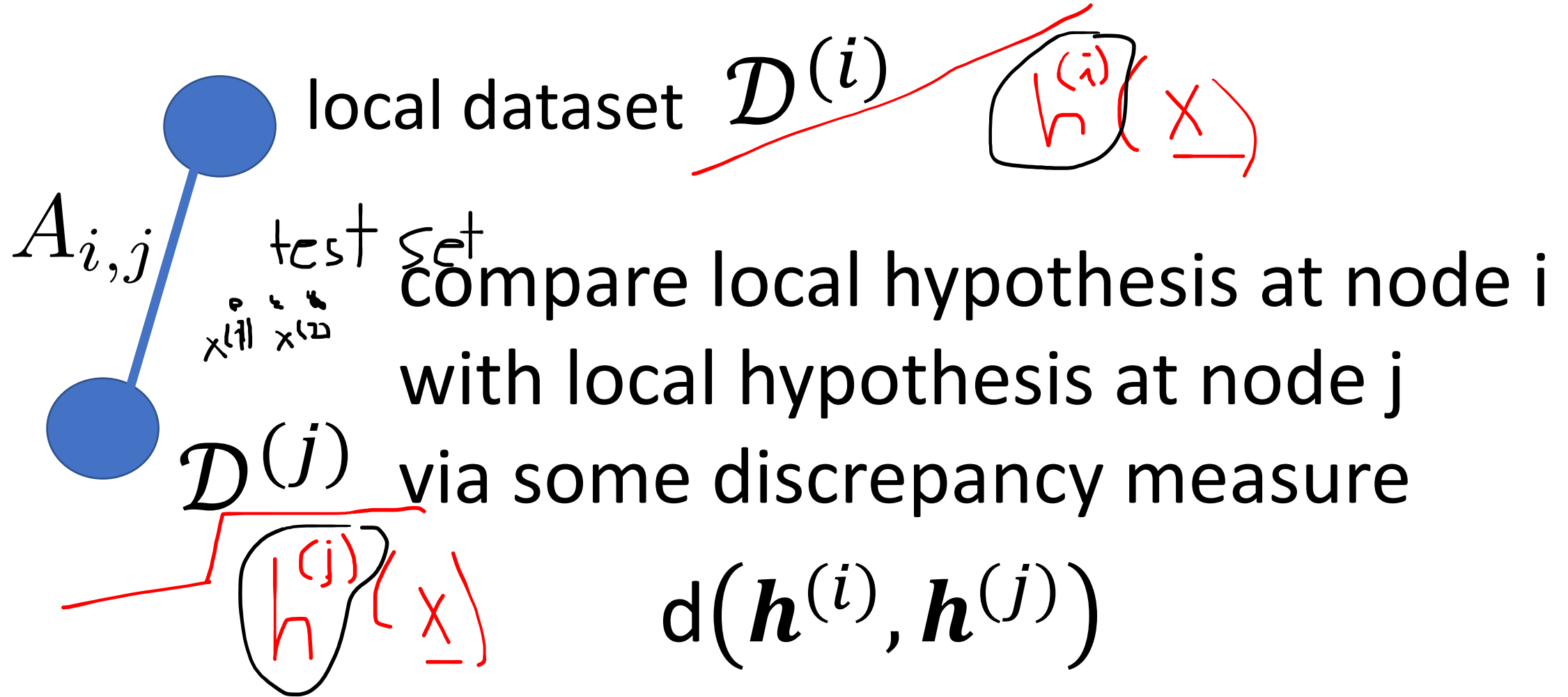
# Choose Edges and Weights

local dataset  $\mathcal{D}^{(i)}$



either choose manually (using your domain expertise) or a graph learning method (see Sec. 7.5 of lecture notes)

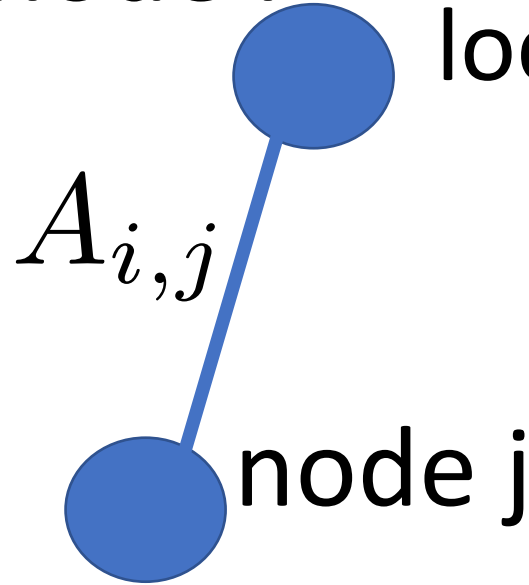
# Discrepancy Measure (Variation)



# Variation of Parametric Models

node i local model param. vec.  $\mathbf{w}^{(i)}$

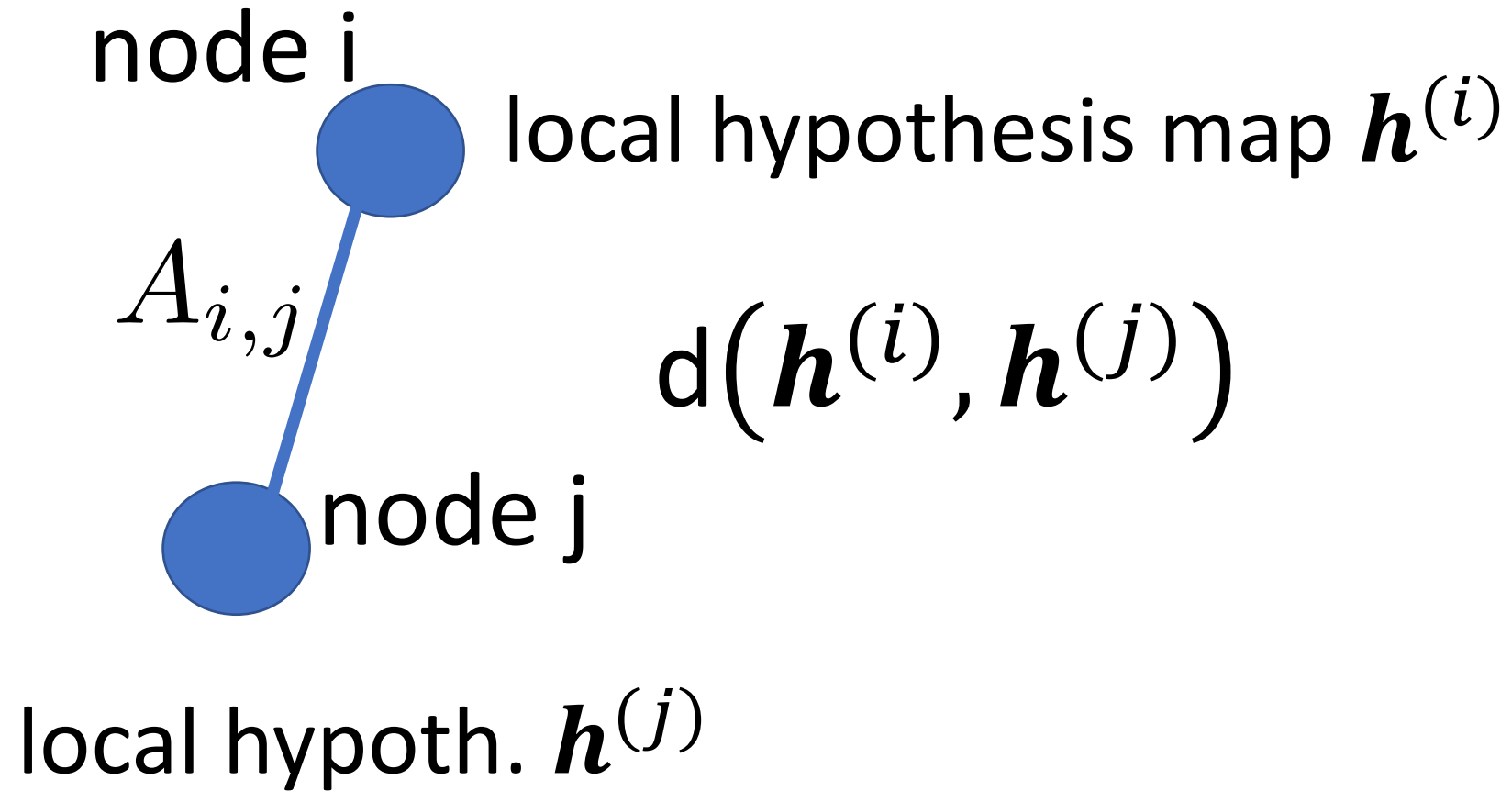
$$h^{(w)}(\underline{x}) = \mathbf{w}^T \underline{x}$$



$$d(\mathbf{h}^{(i)}, \mathbf{h}^{(j)}) = \|\mathbf{w}^{(i)} - \mathbf{w}^{(j)}\|^2$$

local model parameters  $\mathbf{w}^{(j)}$

# Variation of Non-Param. Models





# Putting Together the Pieces

$$\min_w \sum_i L^{(i)}(w^{(i)}) + \lambda \sum_{\{i,j\}} A_{i,j} d(h^{(i)}, h^{(j)})$$

$\left( \sum \frac{(y^{(i)} - w^T x^{(i)})^2}{n(x)} + \alpha \|w\|_2^2 \right)$  Ridge Regression

Fed Relax

Fed SGD

FL Algos obtained by applying (very) well-known optimization methods to solve this GTVMin [1, Sec. 9]

# FL Project Credits and Grades

- extends course from 5 to 10 credits
- project points P2 (max. 80 report/max 20 review)
- basic variant points P1 (max 100)
- grade for 10 credit variant determined from  $\frac{1}{2} ( P1 + P2 )$

**1: 50-59; 2: 60-69; 3: 70-79; 4: 80-89; 5: 90-**

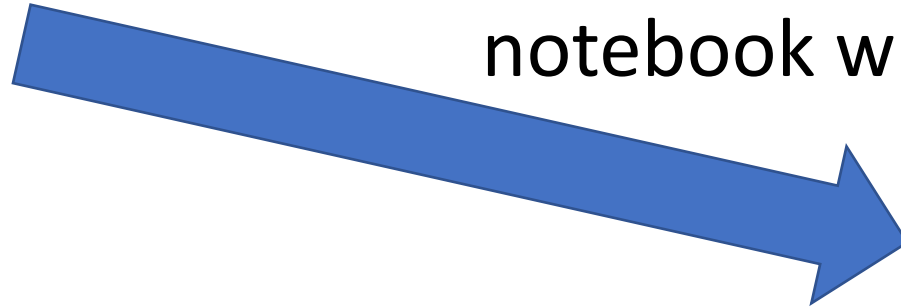
# FL Project Deliverables

- project teams of max. three students
- each student submits project report + notebook
- peer grading during 15 – 31.05.2023
- final submission = project report + python notebook + response to peer reviews

# FL Project Schedule

now

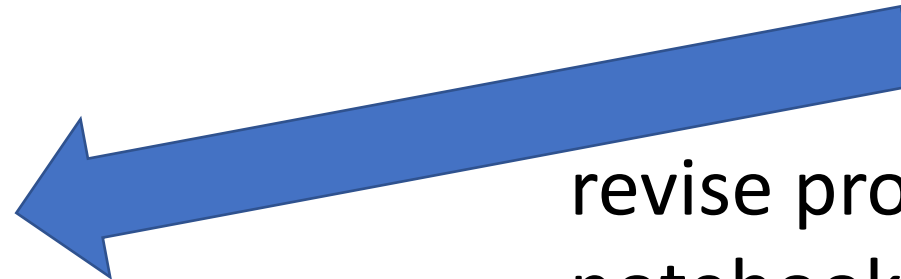
prepare project report +  
notebook with experiments



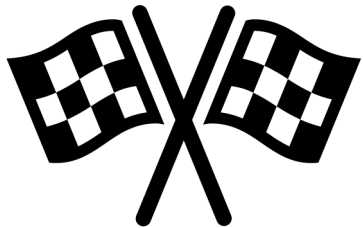
15.05.2023

peer review

31.05.2023



revise project report +  
notebook with experiments  
prepare **response letter**



4/11/23

1. Introduction
2. Problem Formulation
3. Methods
4. Results
5. Conclusion

# 1. Introduction

- describe application domain
- summarize existing work
- outline of paper

## 2. Problem Formulation

provide qualitative description of local datasets:

- what are datapoints? [2, Ch. 2]
- what is the quantity of interest (label)? [2, Ch. 2]
- what is the data source ?
- intrinsic similarity betw. local datas.? [1, Sec. 6]

# 3. Methods

- how data split into train/val/test set? [2, Ch. 6]
- which local models/local loss and why those?
- which discrepancy measure ? [1, Sec. 7.1, 7.4]
- which FL algorithm used ? [1, Sec. 9]
- how did you validate trained models? [2, Ch. 6]



# 4. Results

- report train/val errors for each local model
- diagnose FL algorithm [2, Ch. 6.6.]
- final choice for edge weights, GTVMin param.  $\lambda$
- final chosen local models and their test errors?

# 5. Conclusion

- recap your findings during the project work
- ponder about limitations and possible improvements

# References

- [1] AJ, “Lecture Notes CS-E4740”, 2023
- [2] AJ, “Machine Learning: The Basics,”, Springer, 2023

# Peer Review Questions

draft for you to comment: [click here](#)

revise report and notebook  
based on peer review

# Final Submission:

- revised report (pdf)
- revised notebook with experiments (ipynb)
- response letter that explains how you used the peer review ([sample](#))

Thank you for  
your attention!