



Dafne

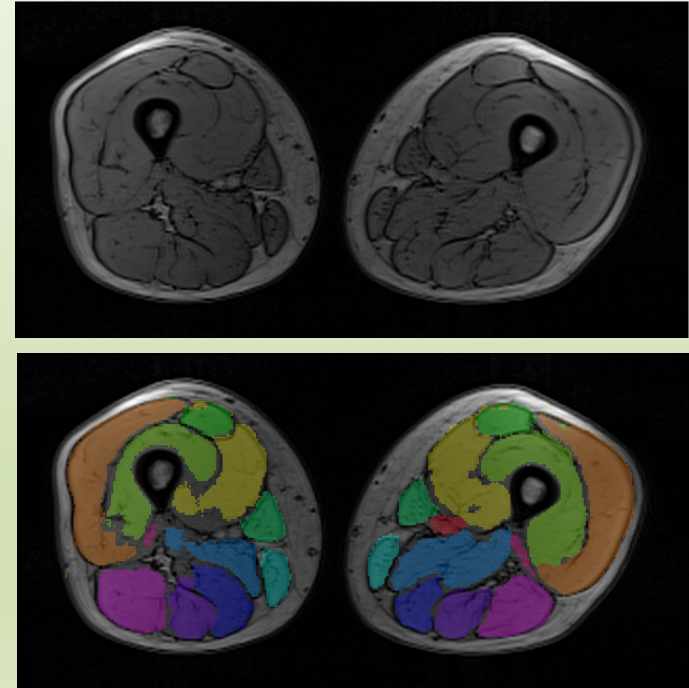
Deep Anatomical Federated Network

Collaborative lifelong learning for MR image segmentation with Dafne: a reproducible research project

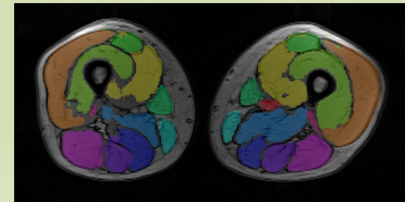
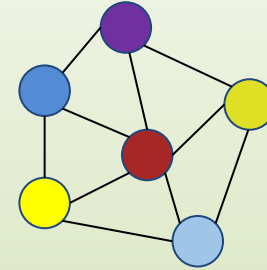
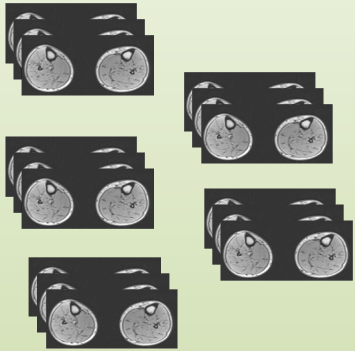
Santini F, Wasserthal J, Agosti A, Pichiecchio A

Muscle segmentation

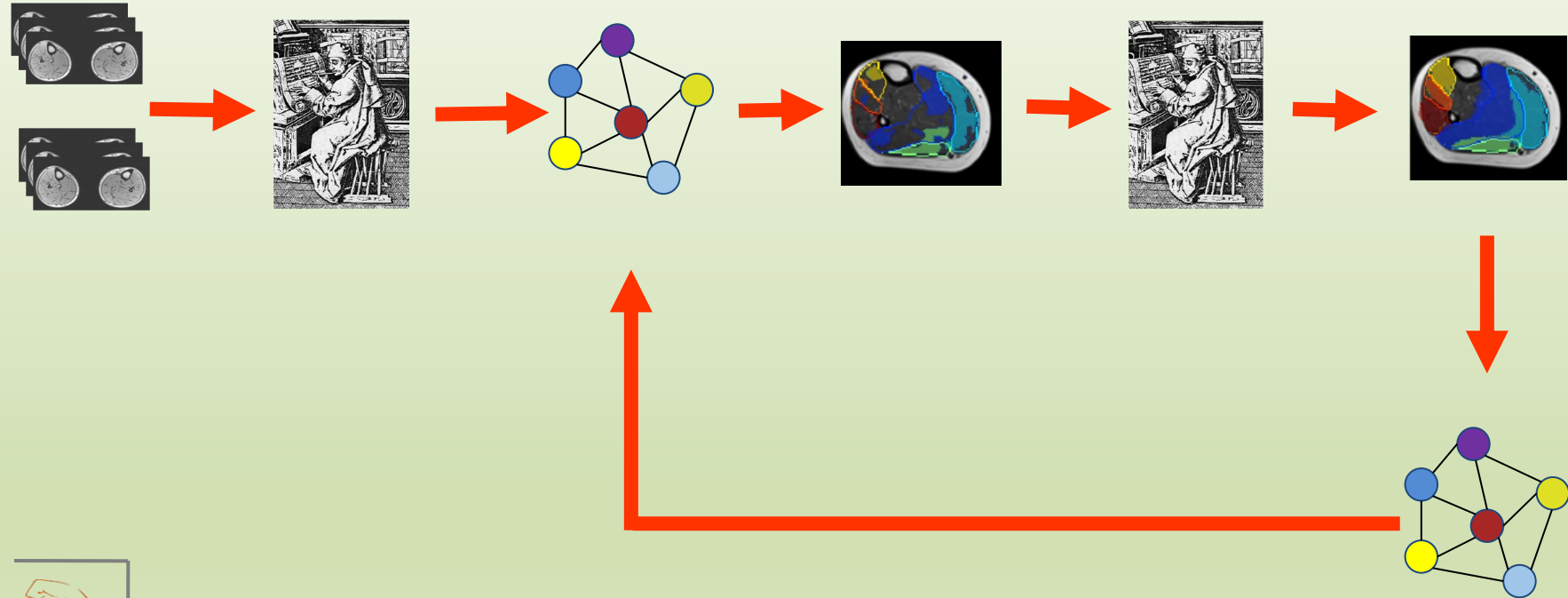
- Quantitative methods rely on segmentation
- Time consuming to do manually
- Complex to automate
 - Deformable geometry
 - Different appearance
 - Rare diseases
 - Multiple contrasts/parameters
- Deep learning is the current standard



Typical segmentation workflow



Incremental learning





Here comes Dafne *

- Dafne is free software for muscle segmentation based on continuous collaborative learning.

- **Thigh and Leg models included**

- **Federated learning**

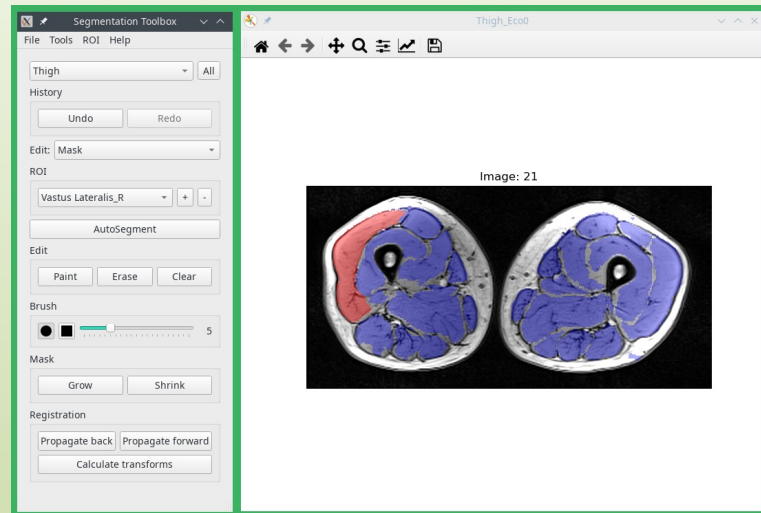
- It collects improvements from all users
 - It preserves data privacy!

- **Continuous incremental learning**

- It learns from your own expertise, even from few examples!

- It has an easy **user interface**

- You always check the segmentation before exporting.

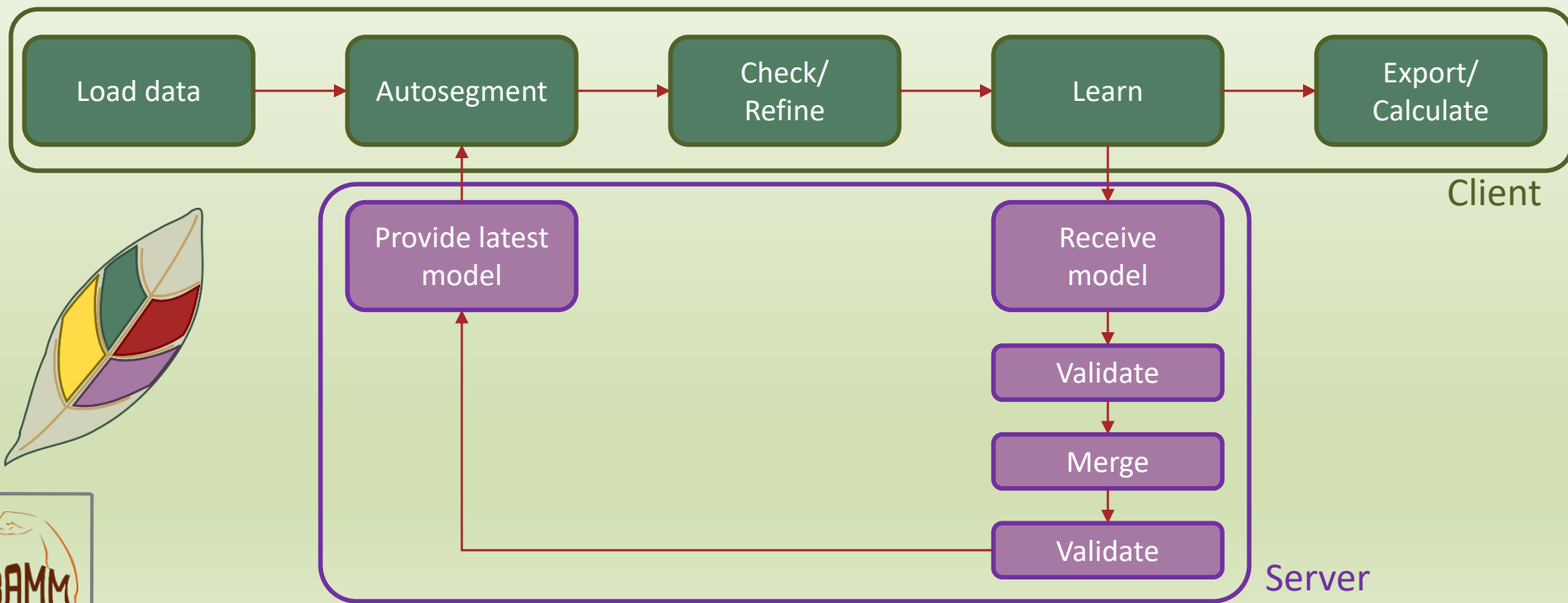


- Get it at <https://dafne.network/>

* Dafne is the Greek name for bay leaf
– and also a girls' name



Dafne Workflow



Does it work?

- We validated it on
 - 38 local datasets, and
 - T1-w images of the calf
 - 18 months of usage statistics!
 - Dice scores collected for thigh and calf

Jun 2021

1

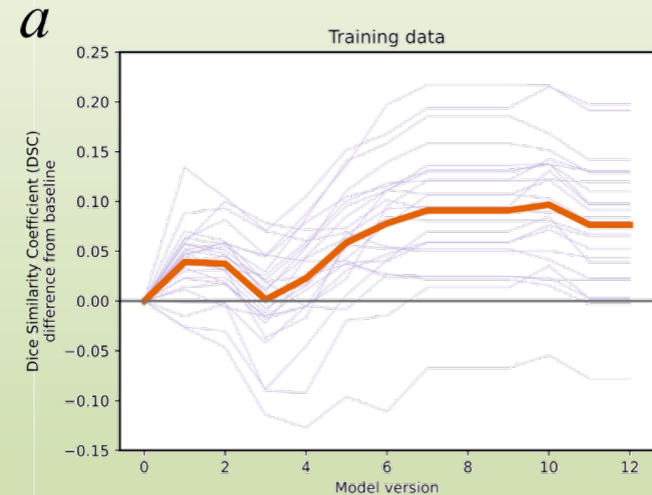
Dec 2022

31



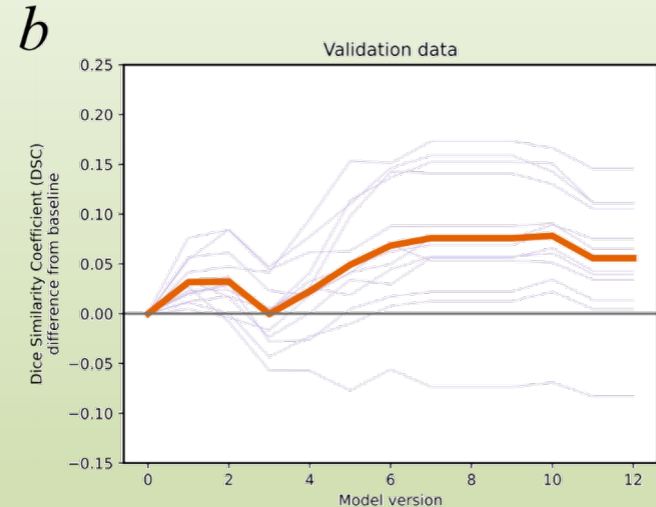
Local validation: training

- 25 datasets used for incremental learning
 - Adaptation to the contrast
- Significant linear increase
 - 0.009 dice points/epoch
 - $p < 0.001$
 - LMM, random slope



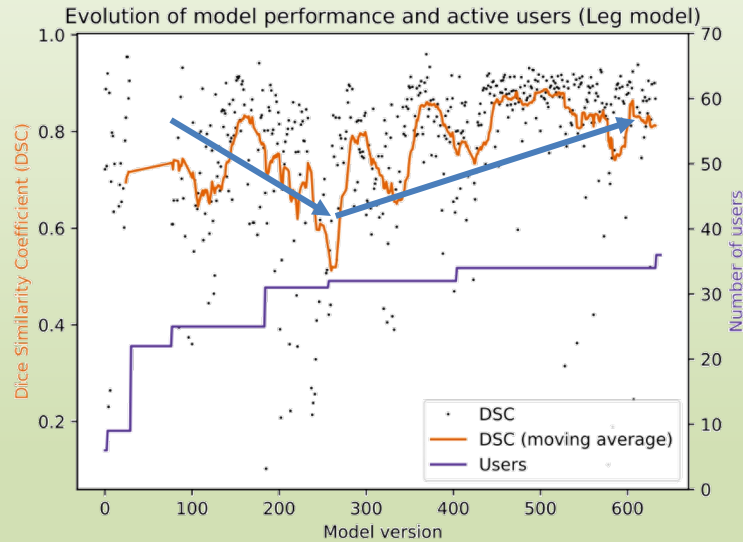
Local validation: validation

- 13 datasets used for validation
 - Tested on the model versions from the previous group
- Significant linear increase
 - 0.007 dice points/epoch
 - $p < 0.001$
 - LMM, random slope

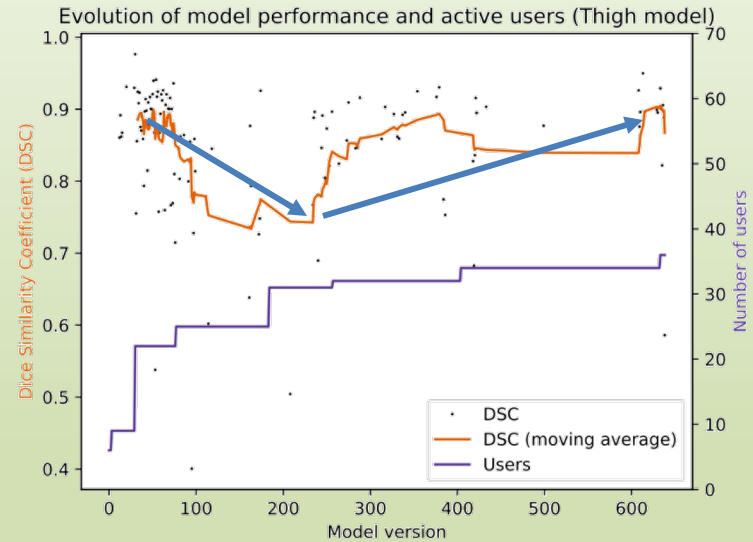


Real-world data

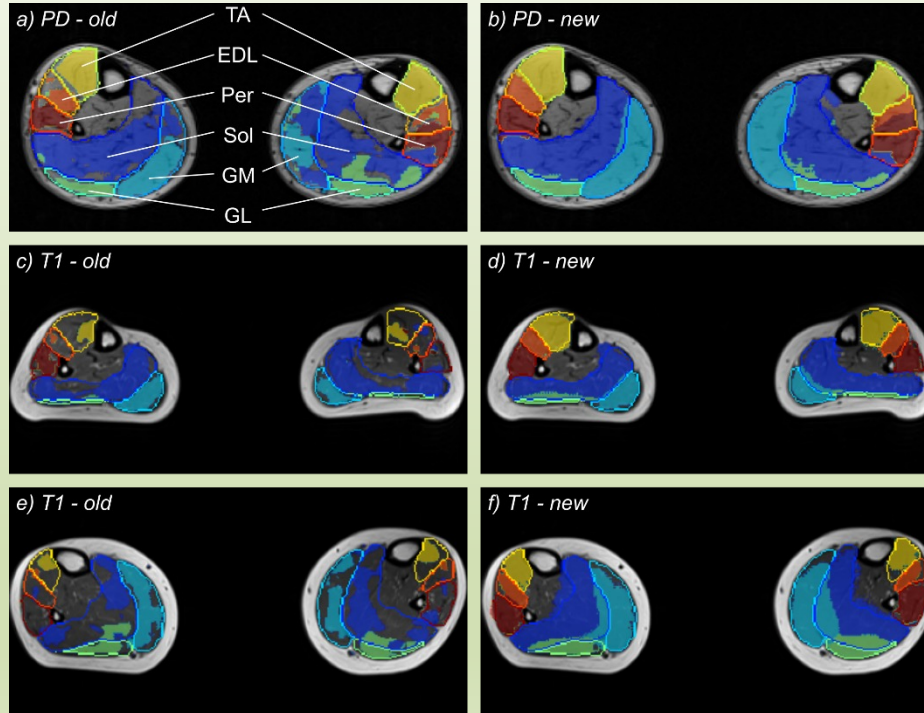
a



b



Segmentation example





Reproducibility

- All development was made public since the beginning
 - Client (GPL): <https://github.com/dafne-imaging/dafne>
 - Server (GPL): <https://github.com/dafne-imaging/dafne-server>
 - Models (GPL): <https://github.com/dafne-imaging/dafne-models>
 - Model interface/common tools (LGPL): <https://github.com/dafne-imaging/dafne-dl>
- All analysis to generate the figures available at – public since the beginning
 - <https://github.com/dafne-imaging/dafne-evaluation>
- Dafne available at
 - <https://dafne.network/>
 - Multiplatform distributions

Evaluation

- Jupyter notebook that produces the images and the statistics
 - Includes data
 - Timestamped (Zenodo)
- Paper on arXiv: <https://arxiv.org/abs/2302.06352>

```
In [22]: # linear mixed model with zero intercept and random slope depending on the dataset
def fit_lm(data_frame):
    print('Important! We are using the relative dice score *1000 to avoid convergence warnings! Rescale the re:
    data_frame['relative_dice_times_1000'] = data_frame['relative_dice']*1000
    md = smf.mixedlm('relative_dice_times_1000 ~ 0 + model_id_rank', data_frame, groups=data_frame['name'], re=
    return md.fit()


In [23]: fit_lm(dice_scores_training).summary()
```

Important! We are using the relative dice score *1000 to avoid convergence warnings! Rescale the results!

Out[23]:

Model:	MixedLM	Dependent Variable:	relative_dice_times_1000
No. Observations:	325	Method:	REML
No. Groups:	25	Scale:	1115.0666
Min. group size:	13	Log-Likelihood:	-1641.2980
Max. group size:	13	Converged:	Yes
Mean group size:	13.0		

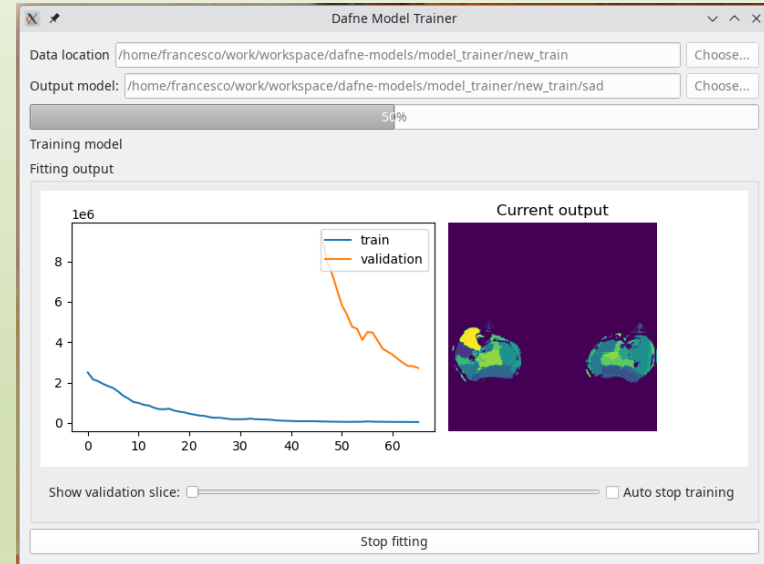
	Coef.	Std.Err.	z	P> z	[0.025	0.975]
model_id_rank	9.050	1.388	6.522	0.000	6.330	11.770
model_id_rank Var	46.427	0.433				

 [launch](#)  [binder](#) DOI [10.5281/zenodo.7629712](https://doi.org/10.5281/zenodo.7629712)

Evaluation repository for the Dafne project

Latest development

- Generic model trainer
- Testing it of kidney images
- Much room for improvement
 - Transfer learning
 - New architectures
 - Transformer (Kanishka)
 - 3D models





Acknowledgment

Our collaborators:

Policlinico Gemelli – Rome

- Giorgio Tasca
- Mauro Monforte
- Enzo Ricci

Fondazione Mondino – Pavia

- Anna Pichiecchio
- Francesca Solazzo
- Matteo Paoletti

Stanford

- Arjun Desai

Peking Union Medical College Beijing

- Fengdan Wang

Siemens China

- Jinxia Zhu

The BAMM Group:

- Xenia Deligianni
- Jakob Wasserthal
- Claudia Weidensteiner
- Tugba Akinci D'Antonoli

Active users who helped improving the model during the Segmenta-thon:

Hermien Kan, Kevin Keene, Christoph Stuprich, Claudia Weidensteiner, Giulia Manco, Valentina Mazzoli, Arjun Desai

