

# Federated Learning on Non-IID Data

Presenter:

Xin Fan

2021/09/22

# Outline

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- ◆ Problem Setup
- ◆ Main Approaches to Handling Non-IID Data
  - ◆ Data based approach
  - ◆ Algorithm based approach
  - ◆ System based approach
- ◆ Discussion





# Outline

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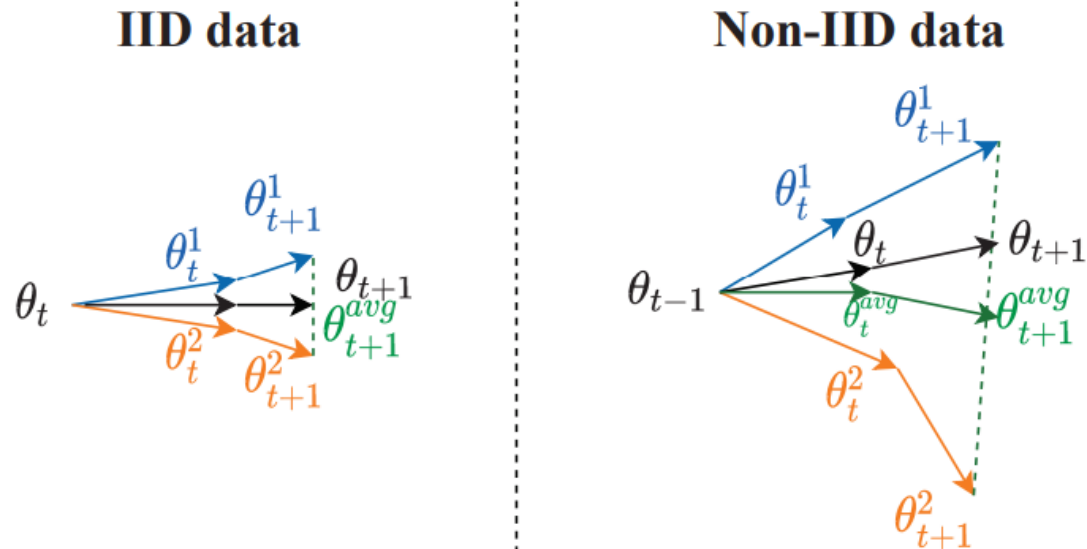
- ◆ **Problem Setup**
- ◆ **Main Approaches to Handling Non-IID Data**
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  - ◆ Algorithm based approach
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# Problem Setup

- FedAvg
- Non-IID federated learning

	bird	deer	frog	ship
Client 1				
Client 2				

- An illusion of parametric model divergence



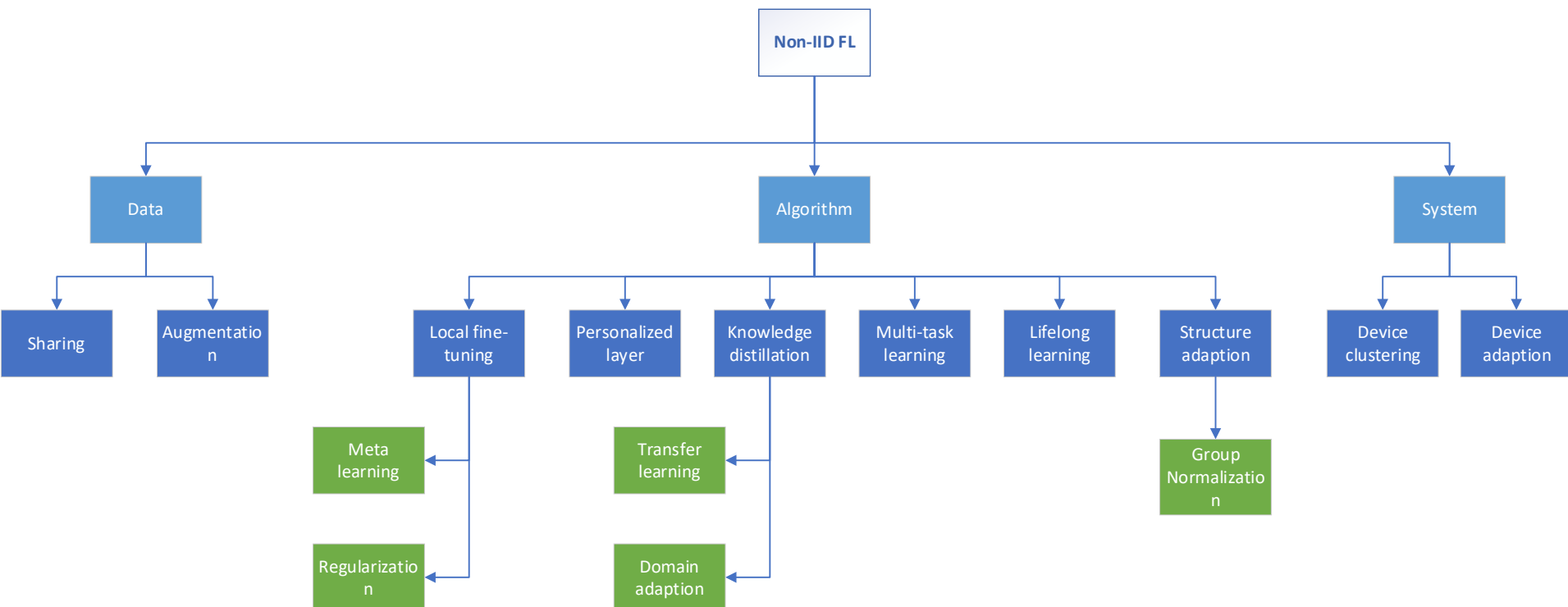
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- ◆ Problem Setup
- ◆ **Main Approaches to Handling Non-IID Data**
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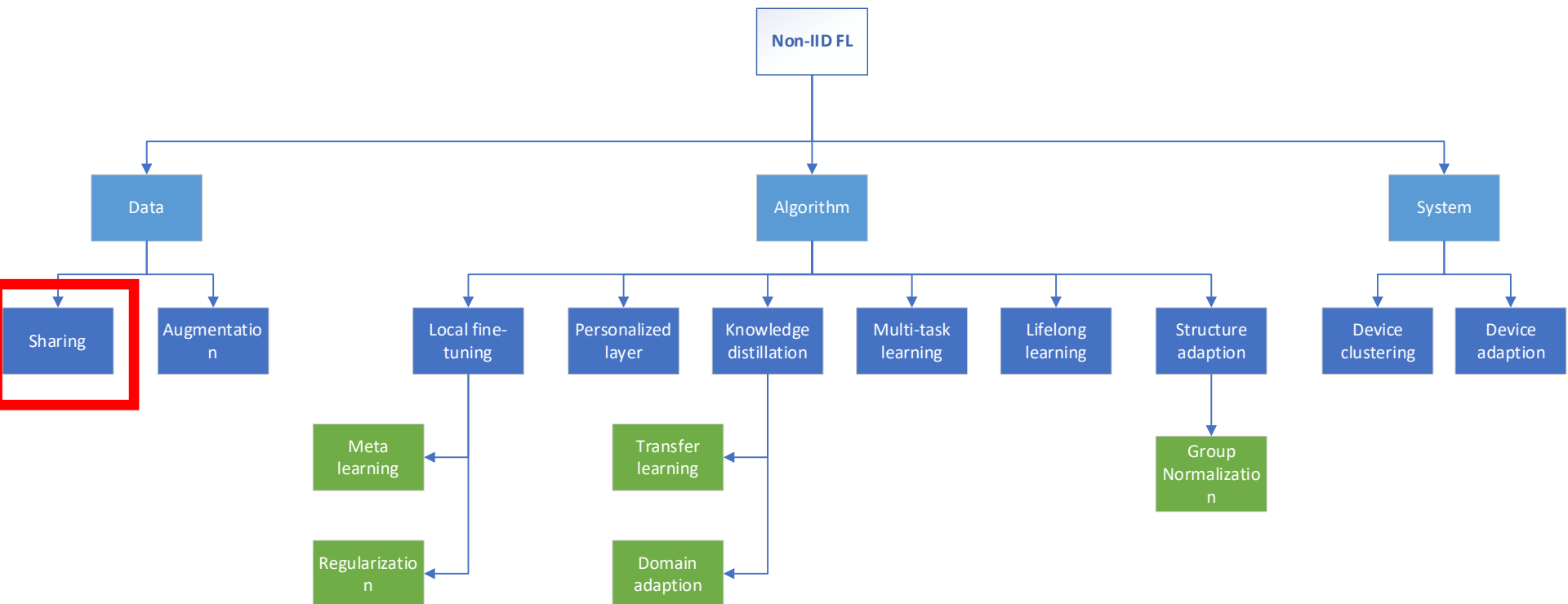
# Main Approaches to Handling Non-IID Data

- A summary of existing approaches to addressing Non-IID data



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# Main Approaches to Handling Non-IID Data

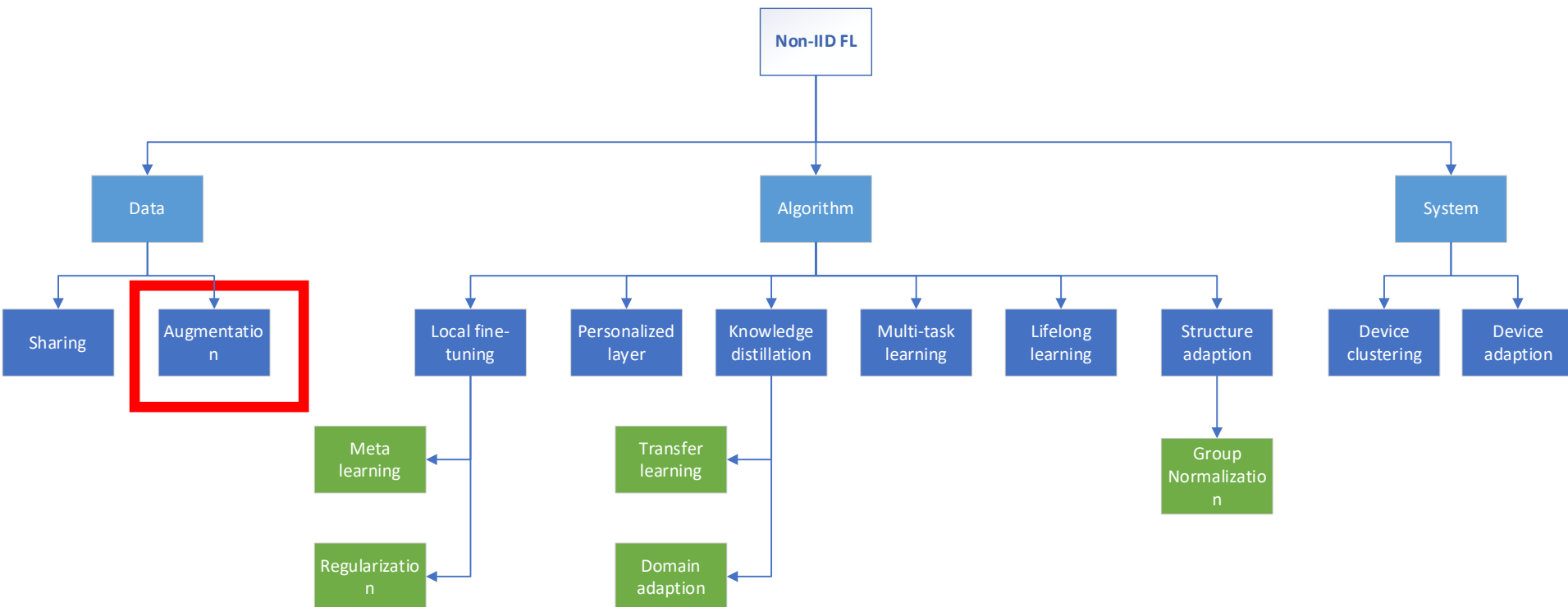
## ■ Data sharing

- [1] Zhao, Y., Li, M., Lai, L., Suda, N., Civin, D., Chandra, V., 2018. Federated learning with non-iid data. arXiv preprint arXiv:1806.00582 .
  - Using a global dataset with a uniform distribution
  - Local models are updated by both training local data and the sharing global dataset
  - Shortcoming: it is hard to get so called uniformly distributed global dataset
  
- [2] Tuor, T., Wang, S., Ko, B.J., Liu, C., Leung, K.K., 2020. Overcoming noisy and irrelevant data in federated learning. arXiv e-prints ,arXiv–2001.
- [3] Yoshida, N., Nishio, T., Morikura, M., Yamamoto, K., Yonetani, R., 2019. Hybrid-fl: Cooperative learning mechanism using non-iid data in wireless networks.
  - Sharing some local data with the server.
  - Shortcoming: privacy issue



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# Main Approaches to Handling Non-IID Data

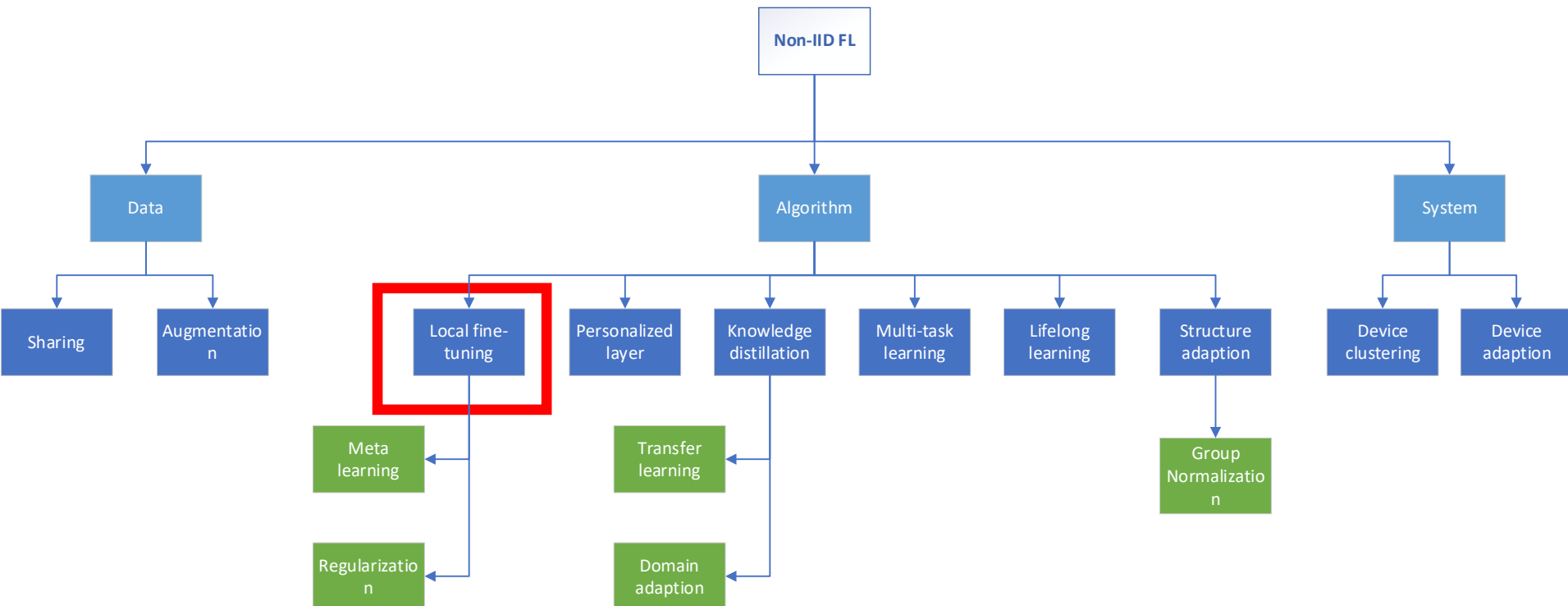
## ■ Data Augmentation

- [4] Duan, M., Liu, D., Chen, X., Tan, Y., Ren, J., Qiao, L., Liang, L., 2019. Astraea: Self-balancing federated learning for improving classification accuracy of mobile deep learning applications, in: 2019 IEEE 37th International Conference on Computer Design (ICCD), IEEE. pp. 246–254.
  - Displace some pixels of the original data samples to get new data samples
  - Both the original local data sample and the augmented data are used to update the local model parameters
- [5] Jeong E, Oh S, Kim H, et al. Communication-efficient on-device machine learning: Federated distillation and augmentation under non-iid private data[J]. arXiv preprint arXiv:1811.11479, 2018.
  - The trained GAN empowers each device to locally reproduce the data samples of all devices, so as to make the training dataset become IID.



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## ■ Local fine-tuning

### ➤ Meta learning

- [6] Fallah A, Mokhtari A, Ozdaglar A. Personalized federated learning with theoretical guarantees: A model-agnostic meta-learning approach[J]. Advances in Neural Information Processing Systems, 2020, 33: 3557-3568.
  - Change the loss function

$$\min_{w \in \mathbb{R}^d} f(w) := \frac{1}{n} \sum_{i=1}^n f_i(w).$$

$$\min_{w \in \mathbb{R}^d} F(w) := \frac{1}{n} \sum_{i=1}^n f_i(w - \alpha \nabla f_i(w)).$$

- The meta-function is defined as  $F_i(w) := f_i(w - \alpha \nabla f_i(w))$

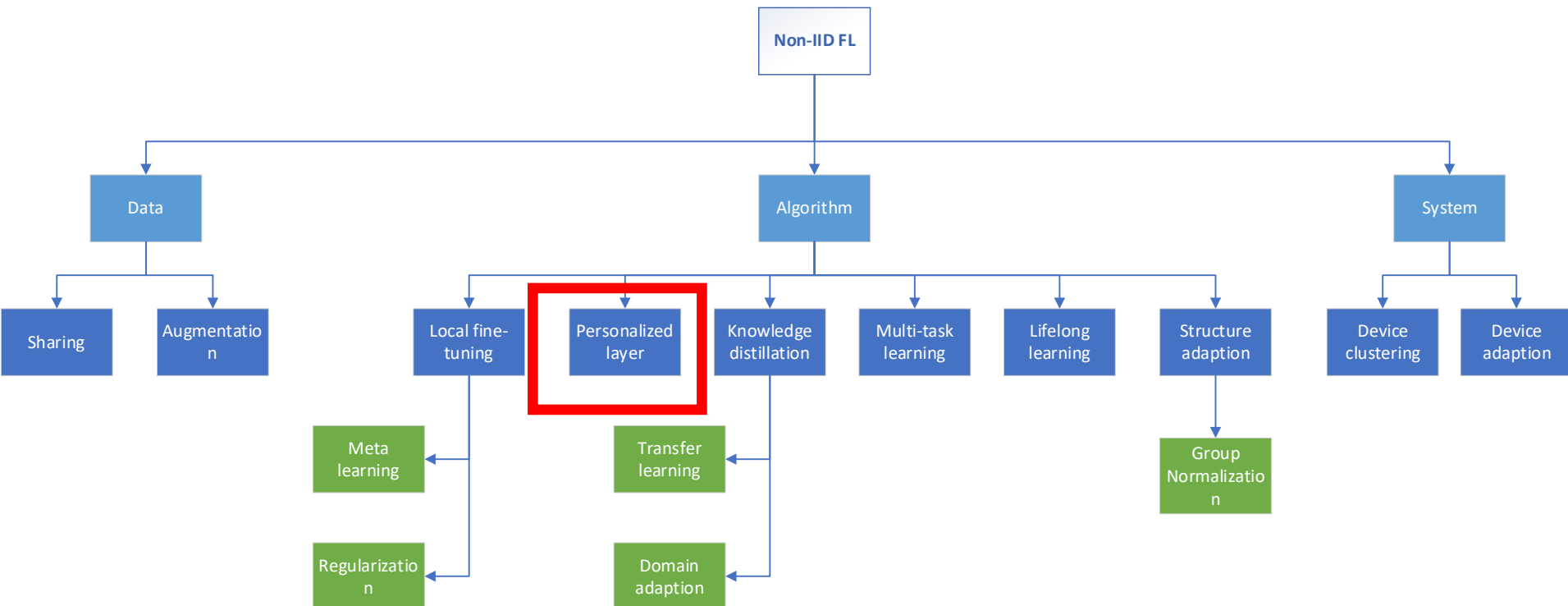
### ➤ Regularization

- [7] Dinh, C.T., Tran, N.H., Nguyen, T.D., 2020. Personalized federated learning with moreau envelopes, in: Advances in Neural Information Processing Systems 33: Annual Conference on Neural Information Processing Systems 2020, NeurIPS 2020, December 6-12, 2020, virtual
- Change the loss function, add a term of L2 norm

$$f_k(\theta_k) = \frac{1}{n_k} \sum_i^{n_k} l(\mathbf{x}_i, y_i; \theta_k) + \frac{\gamma}{2} \|\theta_k - \theta\|^2,$$

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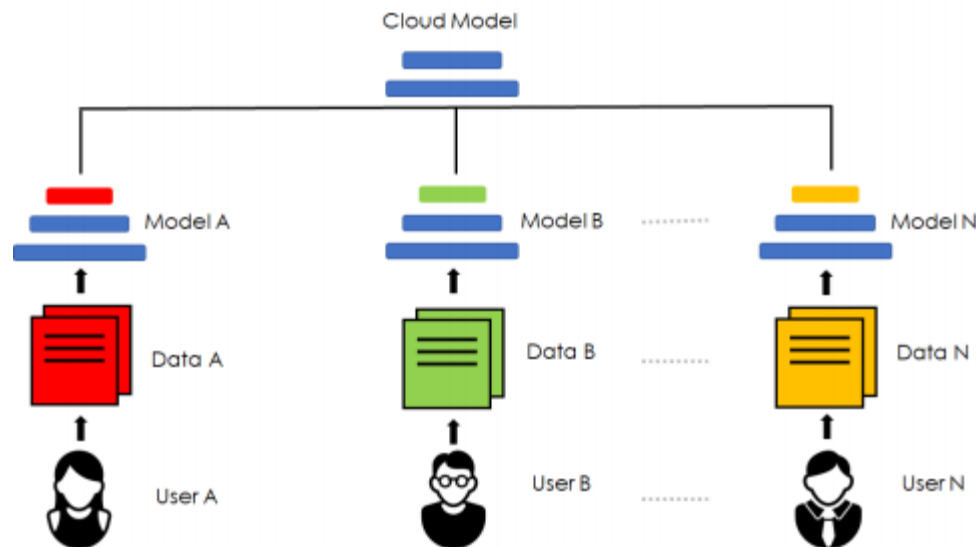
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# Main Approaches to Handling Non-IID Data

## ■ Personalization layer

- [8] Arivazhagan, M.G., Aggarwal, V., Singh, A.K., Choudhary, S., 2019. Federated learning with personalization layers. arXiv preprint arXiv:1912.00818 .
  - The base layers are the shallow layers of the neural network that extracts high-level representations and the personalization layers are the deep layers for classifications



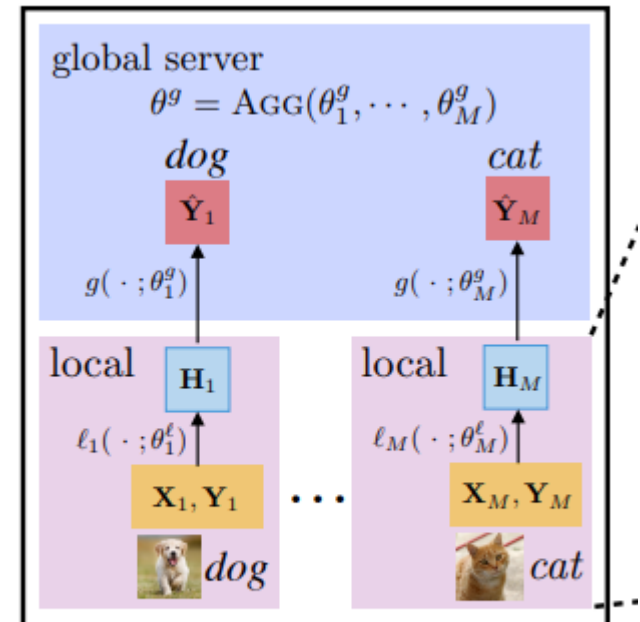
# Main Approaches to Handling Non-IID Data

## ■ Personalization layer

- [9] Liang, P.P., Liu, T., Ziyin, L., Allen, N.B., Auerbach, R.P., Brent, D., Salakhutdinov, R., Morency, L.P., 2020. Think locally, act globally: Federated learning with local and global representations. arXiv preprint arXiv:2001.01523
  - The personalization layers are shallow layers of the neural network and the base layers that are shared with the server are deep layers for class classifications.

## ■ Pros: reduce communication costs

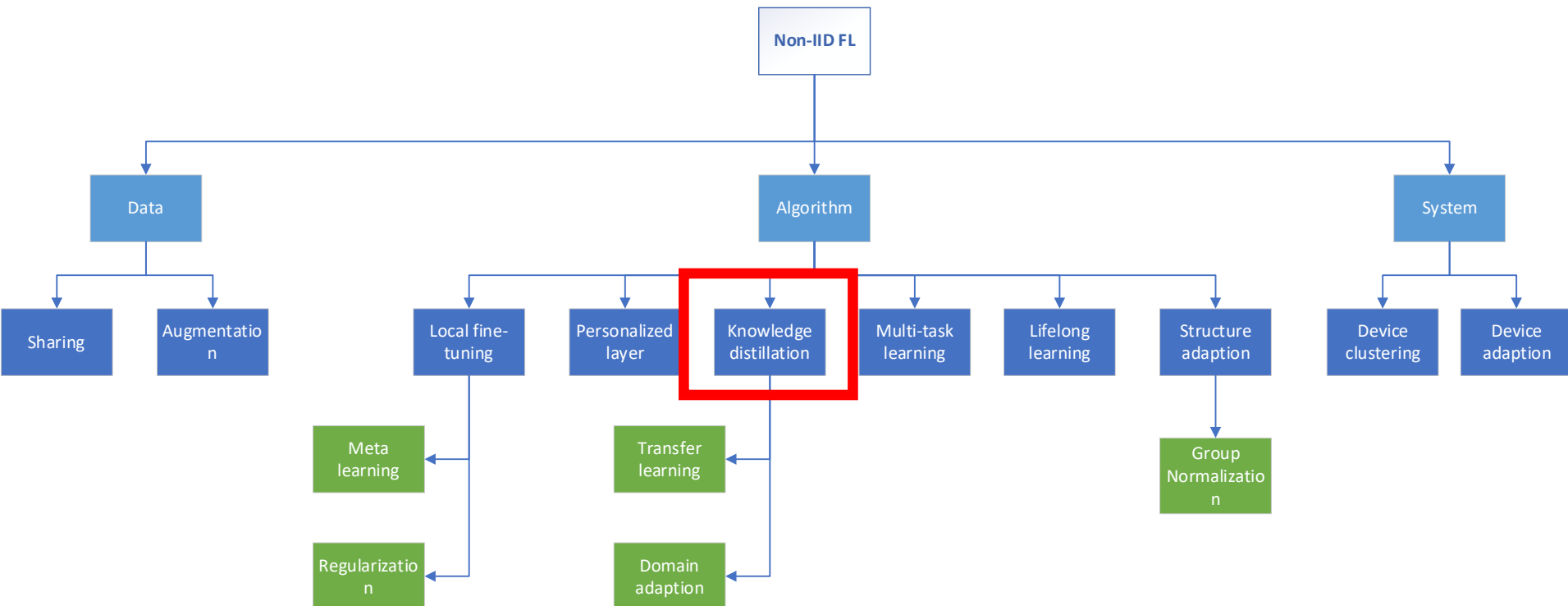
## ■ Cons: storage



Local Global Federated Averaging  
(LG-FedAvg, our proposed method)

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# Main Approaches to Handling Non-IID Data

- Knowledge distillation

- Transfer learning

- Transfer knowledge from the server or other clients to a certain client to improve its performance on unknown heterogeneous data
    - [10] K. Wang, R. Mathews, C. Kiddon, H. Eichner, F. Beaufays, and D. Ramage, “Federated evaluation of on-device personalization,” arXiv preprint arXiv:1910.10252, 2019.
    - Personalization on a well trained global model.

- Domain adaption

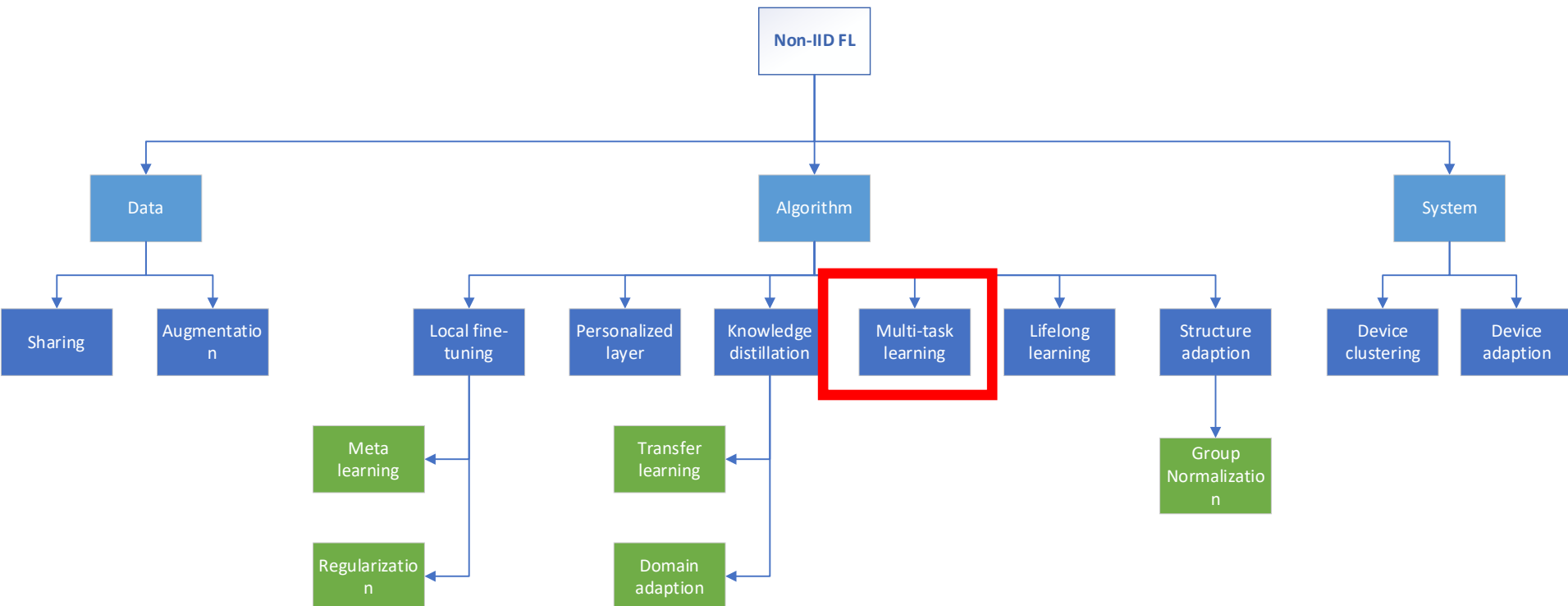
- [11] Peterson D, Kanani P, Marathe V J. Private federated learning with domain adaptation[J]. arXiv preprint arXiv:1912.06733, 2019.

$$\hat{y}_i = \alpha_i(x)M_G(x, \Theta_G) + (1 - \alpha_i(x))M_{P_i}(x, \Theta_{P_i}).$$

$\alpha_i(x) = \sigma(w_i^T \cdot x + b_i)$ , where  $\sigma(x)$  is the sigmoid function, and  $w_i$  and  $b_i$  are learned weights.

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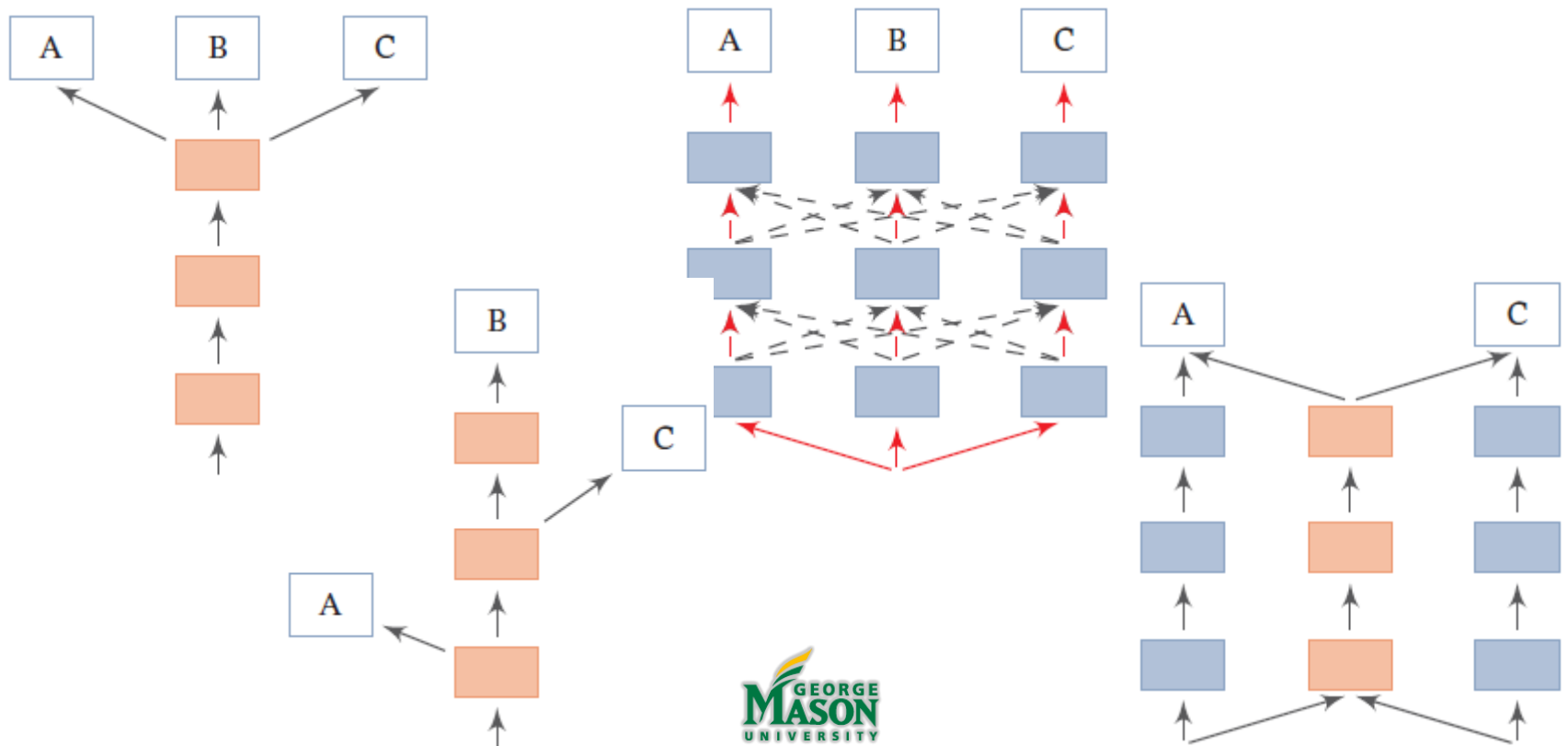
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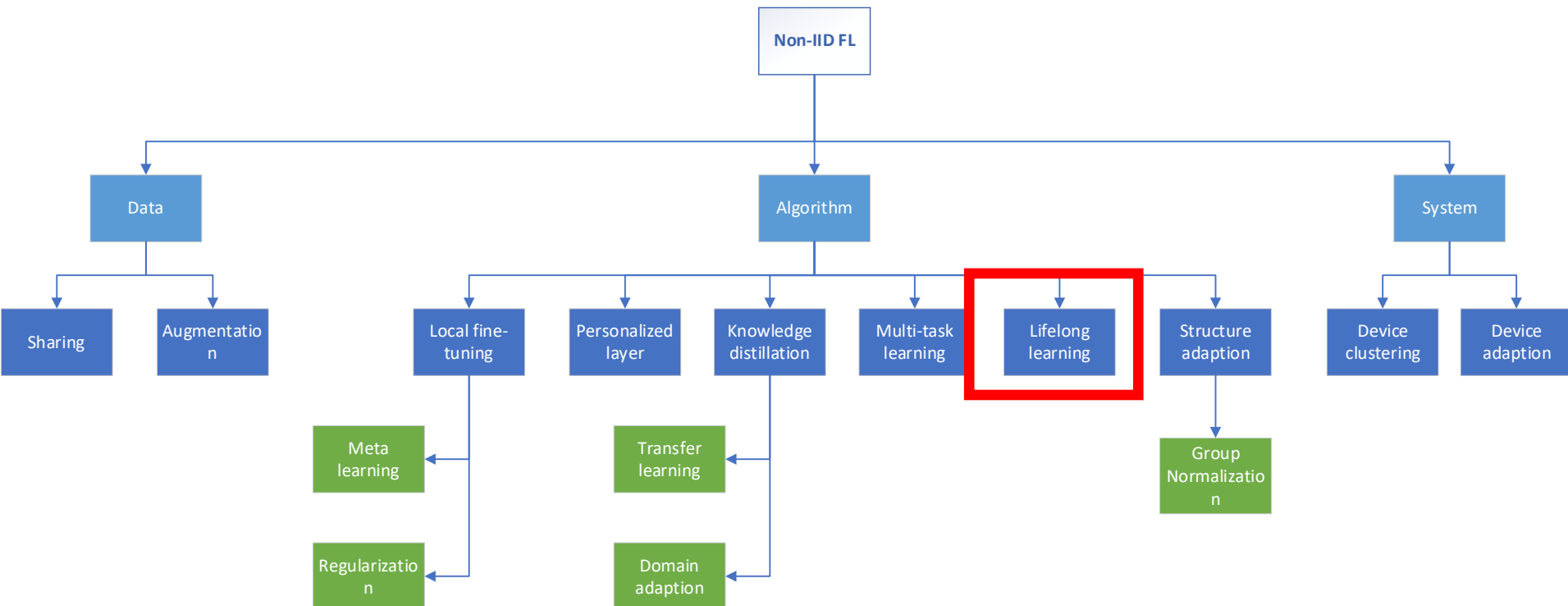
- Multi-task learning

- [12] Caruana, R., 1997. Multitask learning. Machine learning 28, 41–75.



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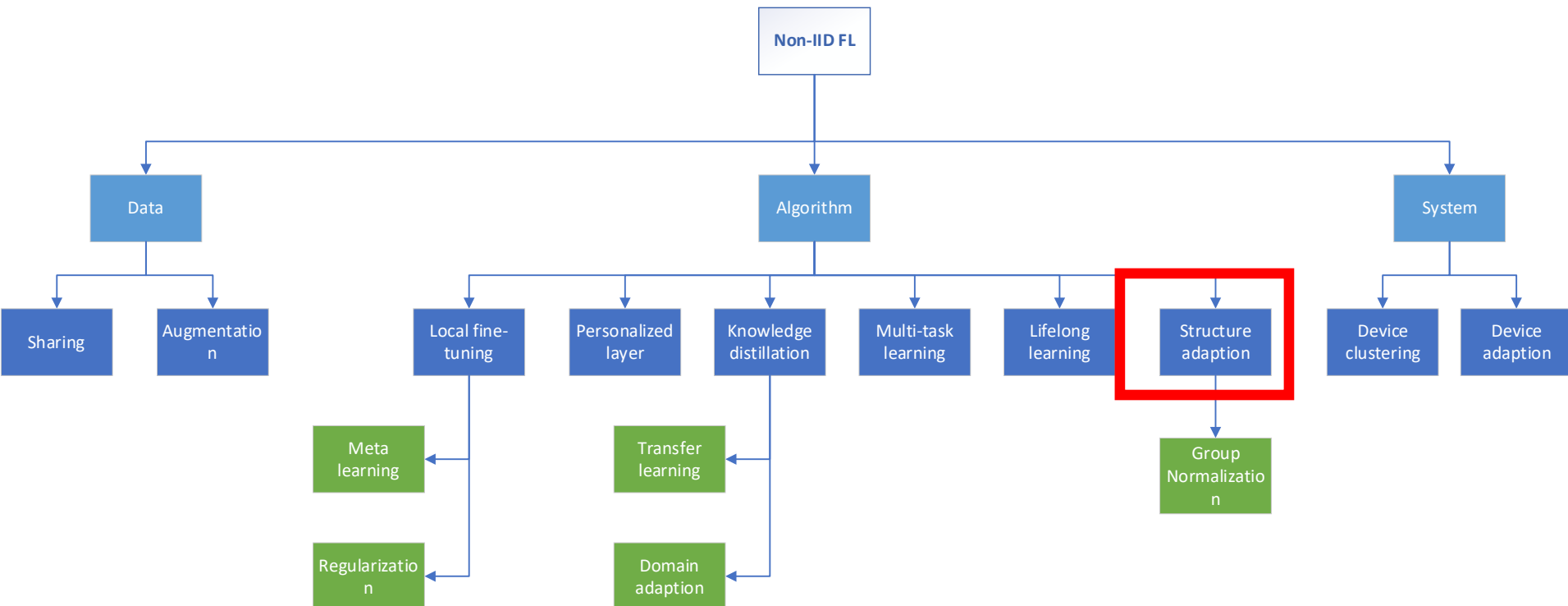
# Main Approaches to Handling Non-IID Data

## ■ Lifelong Learning

- Maintain the model accuracy without forgetting previously learnt tasks
- [13] Kirkpatrick, J., Pascanu, R., Rabinowitz, N., Veness, J., Desjardins, G., Rusu, A.A., Milan, K., Quan, J., Ramalho, T., Grabska-Barwinska, A., et al., 2017. Overcoming catastrophic forgetting in neural networks. Proceedings of the national academy of sciences 114, 3521–3526.
  - The most important parameters for a specific task A are identified.
  - When the model is trained on another task B, the learner will be penalized for changing these parameters

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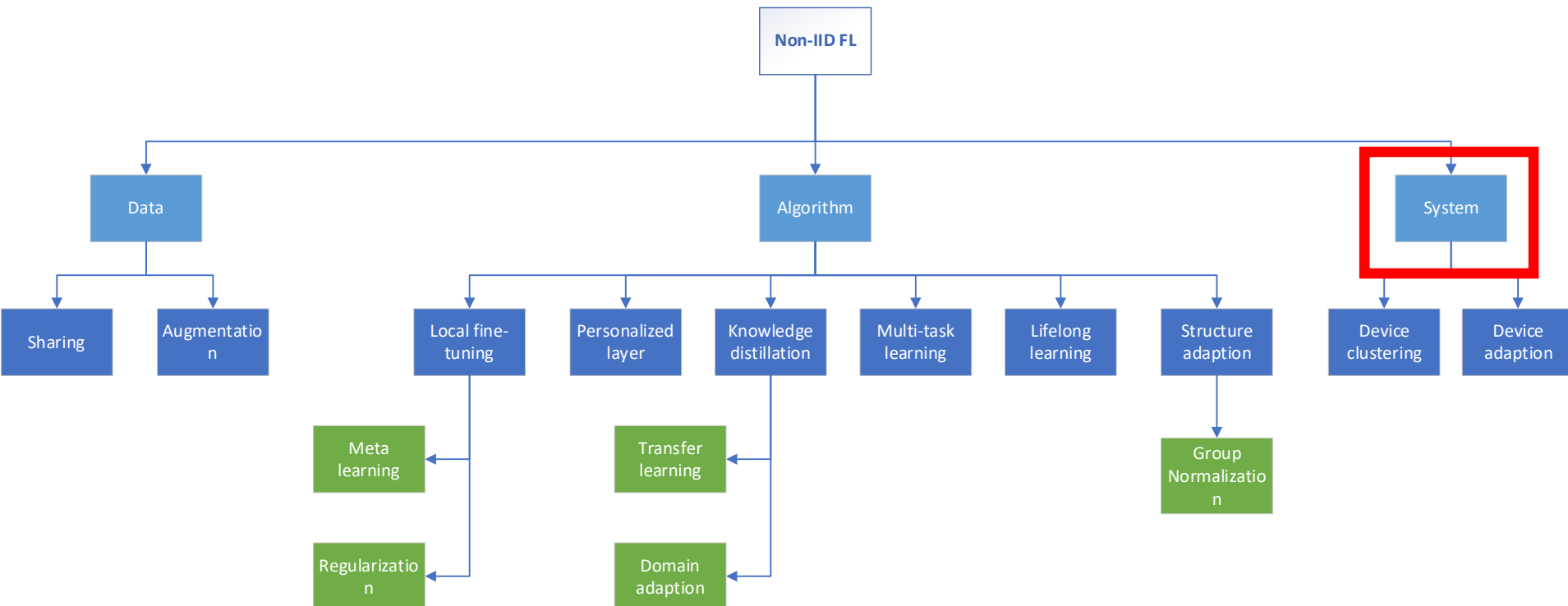
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- Structure adaption

- Group normalization
- [14] Zhang, Z., Yang, Y., Yao, Z., Yan, Y., Gonzalez, J.E., Mahoney, M.W., 2020b. Improving semi-supervised federated learning by reducing the gradient diversity of models. arXiv preprint arXiv:2008.11364 .
  - Add some group normalization layers into model.

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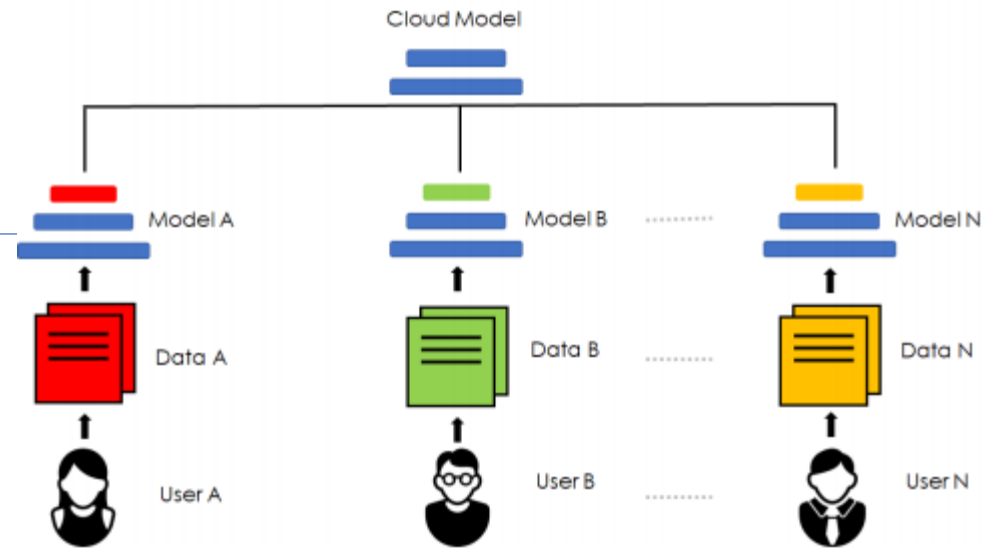
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# Discussion

- Personalization

- Base layers are shared layers[8]
- Deep layers are shared layers[9]



- Which layers should be personalized

- [15] Li X C, Gan L, Zhan D C, et al. Aggregate or Not? Exploring Where to Privatize in DNN Based Federated Learning Under Different Non-IID Scenes[J]. arXiv preprint arXiv:2107.11954, 2021.

