

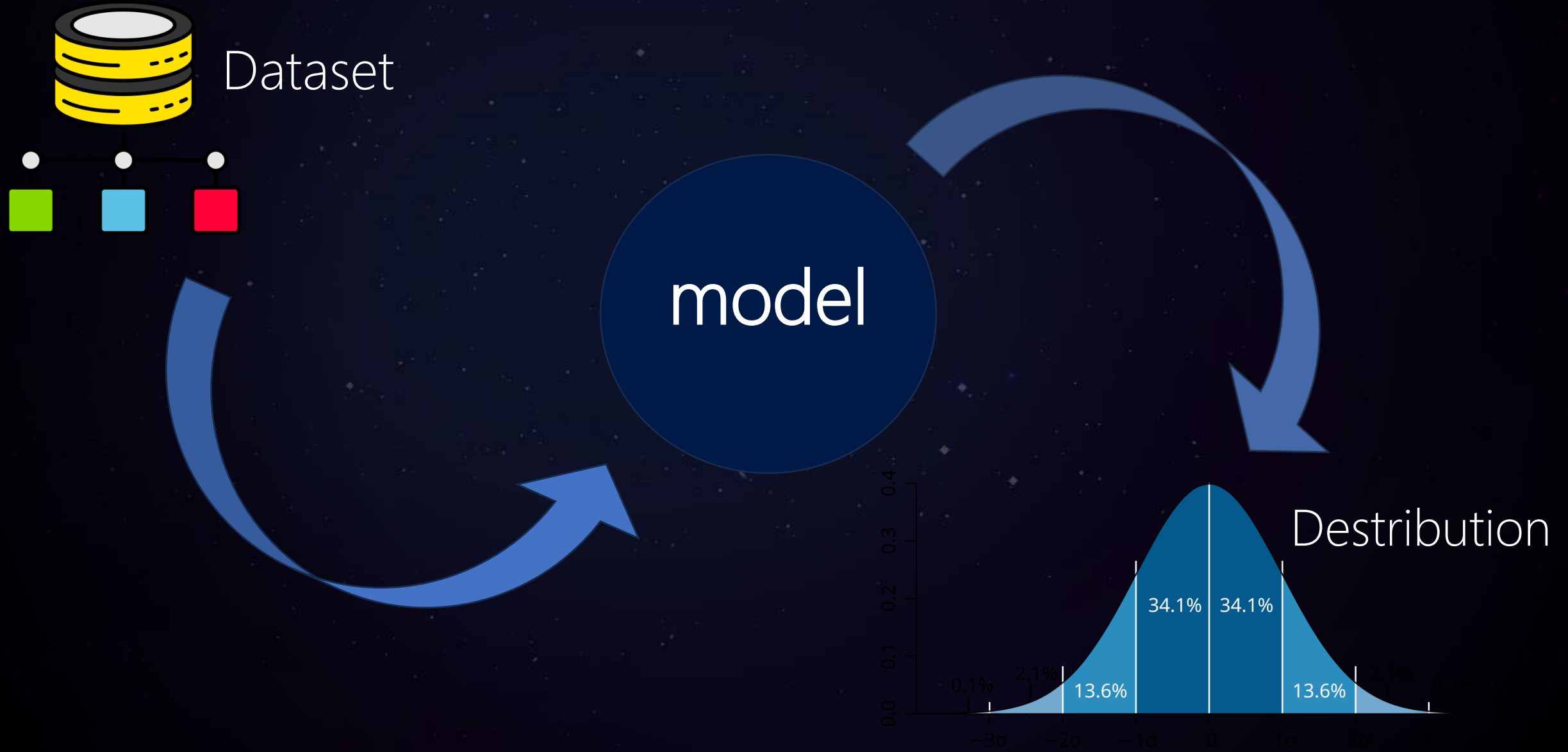
# FEDERATIVE LEARNING

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Sajjad Ranjbar



# CLASSICAL MACHINE LEARNING



# CLASSICAL MACHINE LEARNING

Data is  
confidential



Dataset

The volume of data  
is  
enormous

# CLASSICAL MACHINE LEARNING



Dataset

The volume of data  
is  
**enormous**

A single server must  
compute all processes

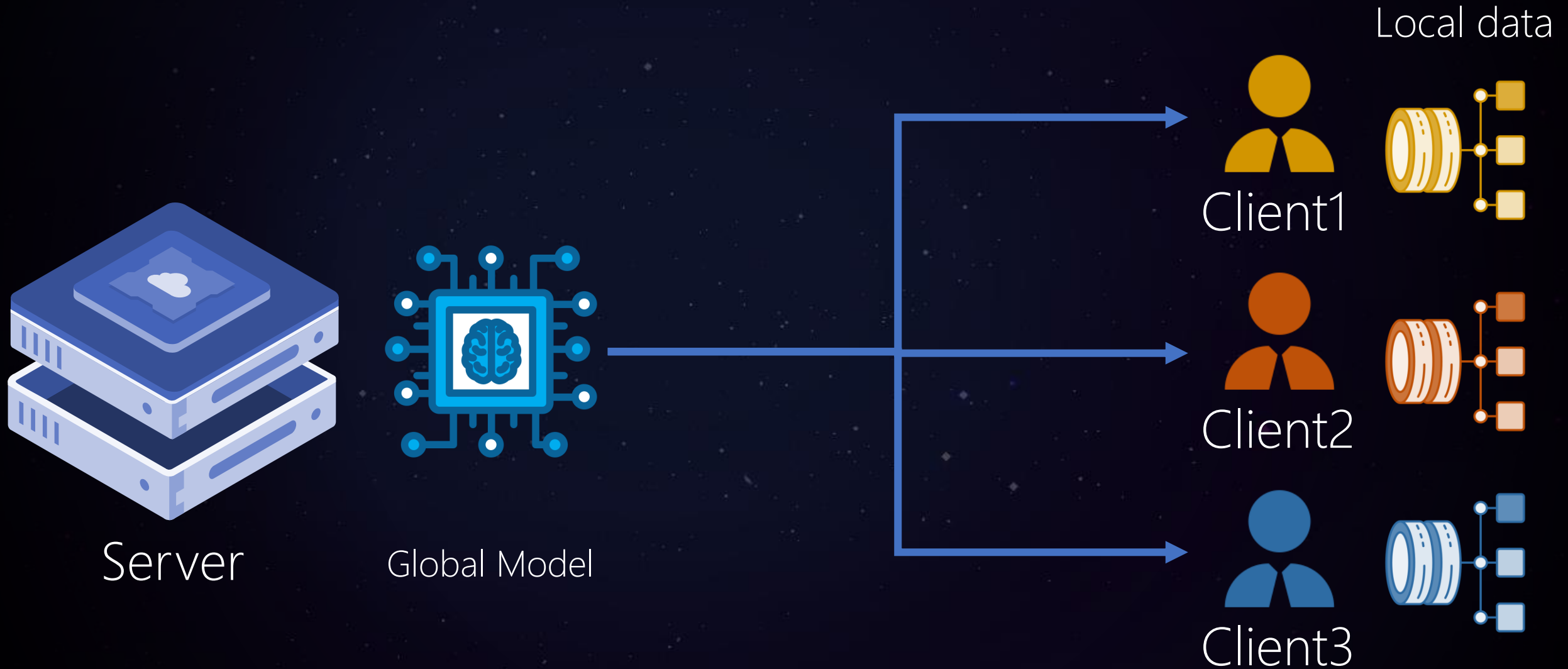
Data is  
**confidential**

Medical information,  
banking transactions, etc.

# FEDERATED LEARNING

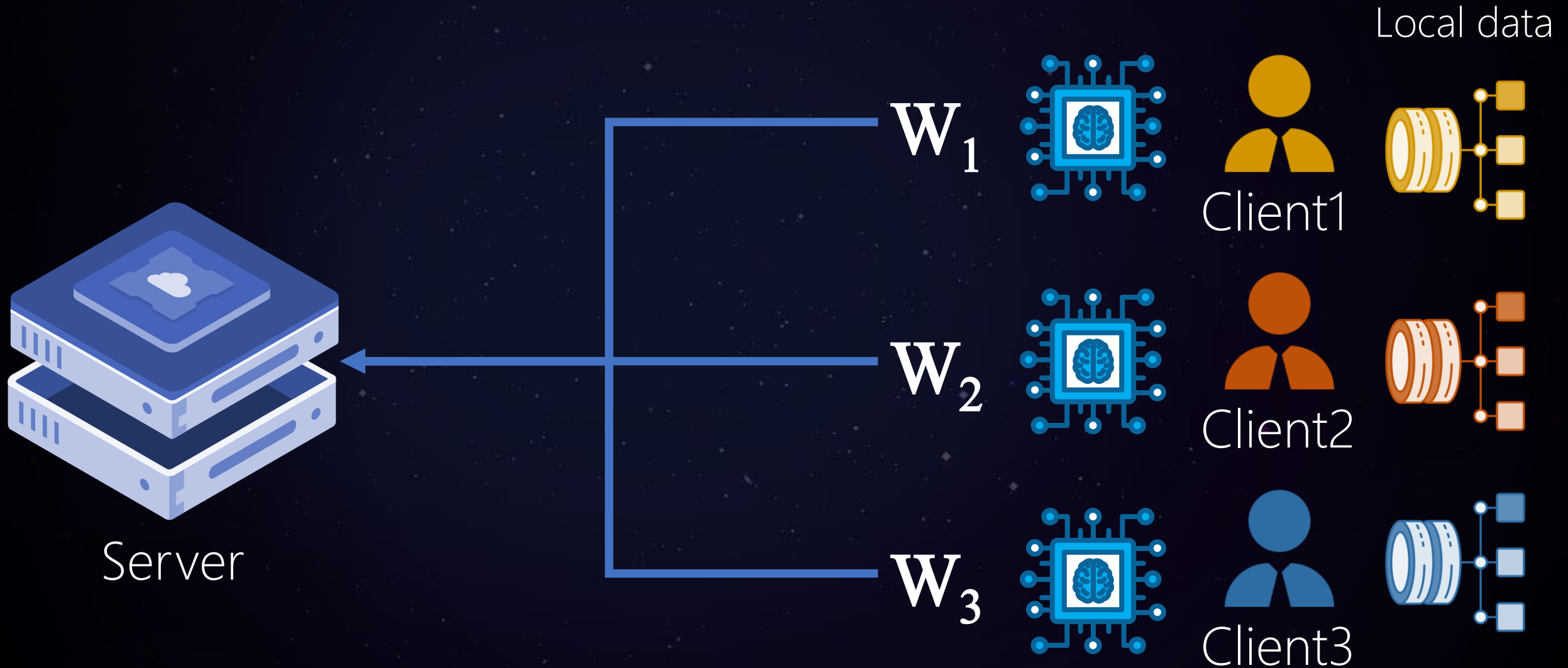
Federated learning is a distributed machine learning approach where multiple devices or entities collaborate to train a shared model without exchanging raw data. This allows for data privacy and security while still enabling collaborative model improvement.

# FEDERATED LEARNING





# FEDERATED LEARNING

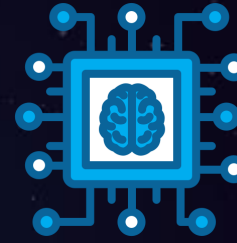


# FEDERATED LEARNING

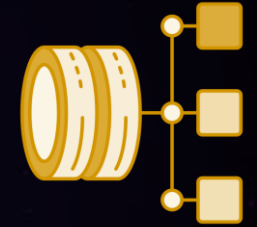


Server

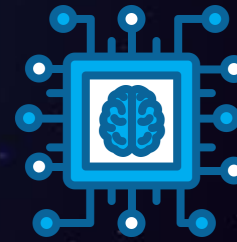
$$W = F(W_1, W_2, W_3)$$



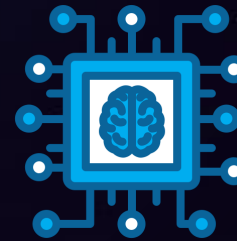
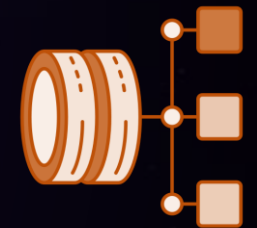
Client1



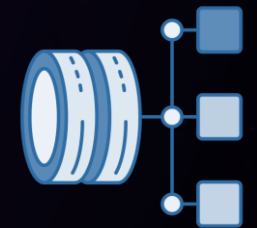
Local data



Client2

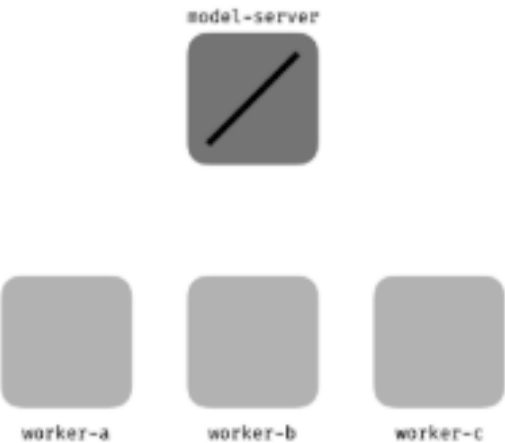
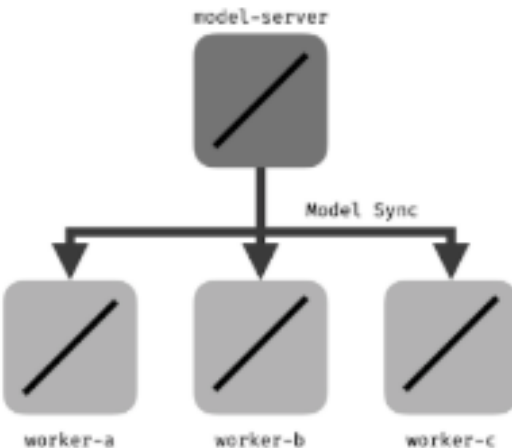
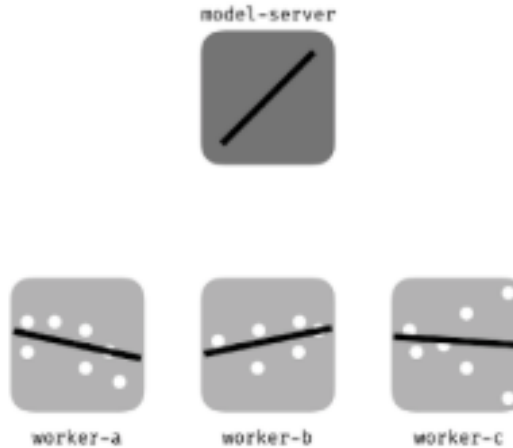
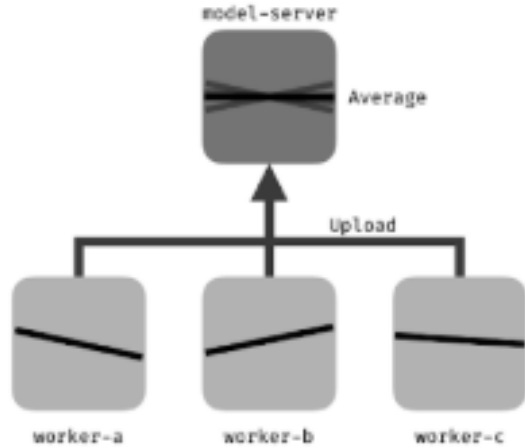


Client3





# FEDERATED LEARNING

Step 1	Step 2	Step 3	Step 4
 <p>model-server</p> <p>worker-a worker-b worker-c</p>	 <p>model-server</p> <p>Model Sync</p> <p>worker-a worker-b worker-c</p>	 <p>model-server</p> <p>worker-a worker-b worker-c</p>	 <p>model-server</p> <p>Average</p> <p>Upload</p> <p>worker-a worker-b worker-c</p>
Central server chooses a statistical model to be trained	Central server transmits the initial model to several nodes	Nodes train the model locally with their own data	Central server pools model results and generate one global mode without accessing any data

# FEDERATED LEARNING

Even in FL, model weights may provide the original data.

**Solution:** Encrypt weights before sending to the server.

**Challenge:** The server must be able to perform aggregation operations on encrypted weights.



Homomorphic Encryption

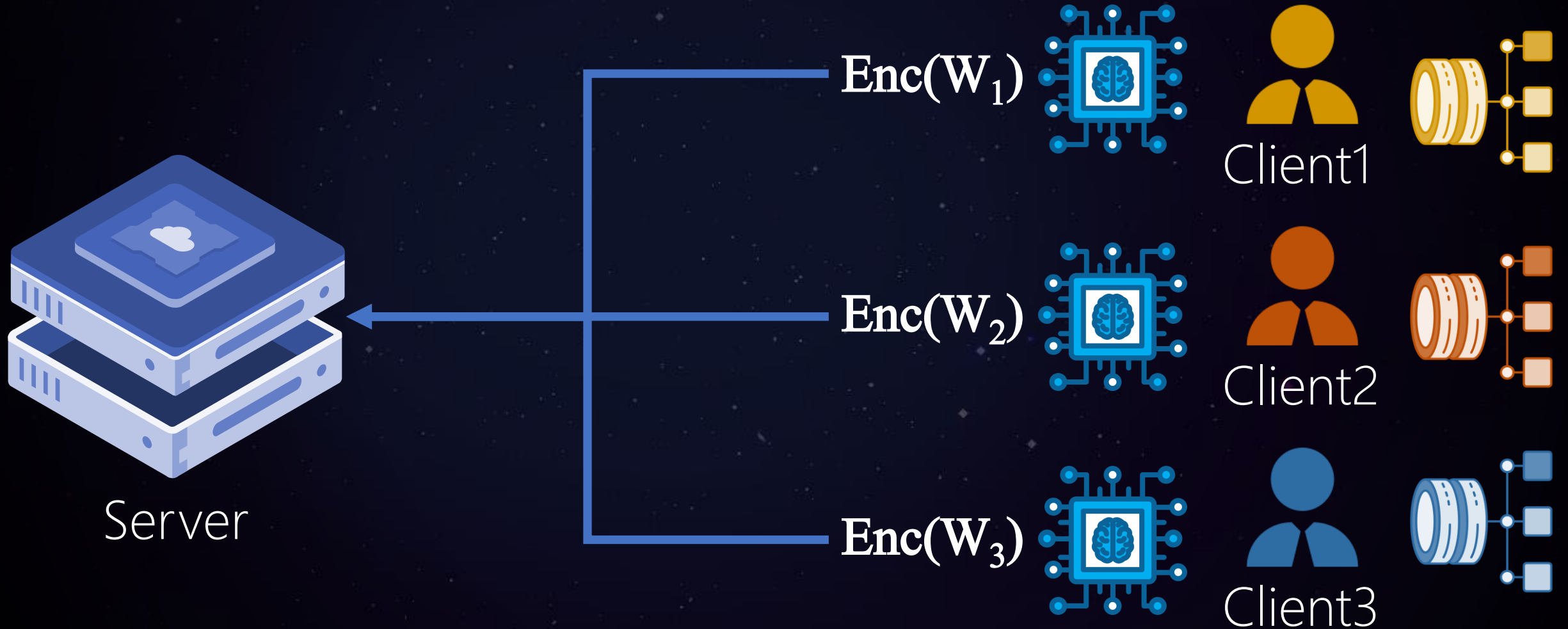
# HOMOMORPHIC ENCRYPTION

A type of encryption that allows calculations to be performed on encrypted data without the need to decrypt it.

$$\text{Enc}(x \oslash y) = \text{Enc}(x) \oslash \text{Enc}(y)$$

# FEDERATED LEARNING

Local data

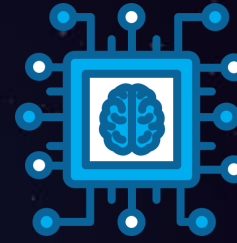


# FEDERATED LEARNING



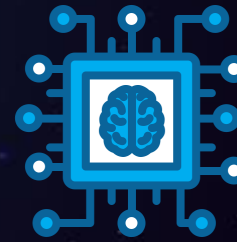
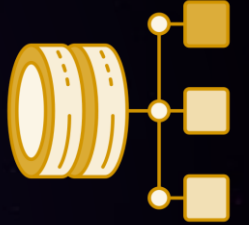
Server

$$W = F(\text{Enc}(W_1), \text{Enc}(W_2), \text{Enc}(W_3))$$

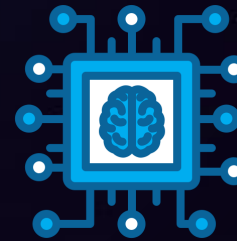
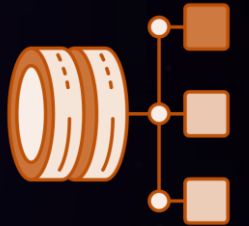


Client1

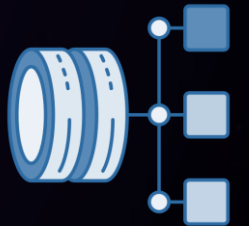
Local data



Client2



Client3



# FEDERATED LEARNING

Real example



id	text	label
1	...	Suicide
2	...	Non-suicide

[Link](#)



# EXAMPLE

Many Classical models have been tested on this dataset. You can see a good example in this [notebook](#). The following methods have been tested on this netbook:

- Naive Bayes (Voting Classifier)
- Random Forest
- Decision Tree
- Gradient Boosting
- XG Boost

# FEDERATED LEARNING BASED ON NAIVE BAYES



Server

Naive  
Bayes

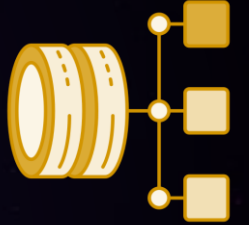
$$W = \frac{1}{20} \sum_{i=1}^{20} w_i$$

$W_1$



Client1

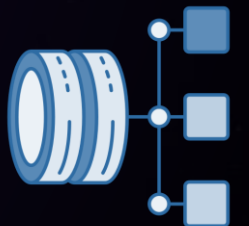
Local data



$W_{20}$



Client20



# RESULTS

	Classical model	Federated model(20 clients)
Train Accuracy	88.02%	-
Test Accuracy	88.06%	88.53
Precision	0.86	0.89
Recall	0.91	0.88
F1-score	0.88	0.89

Thank you so much