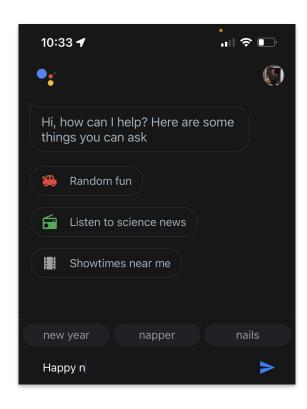
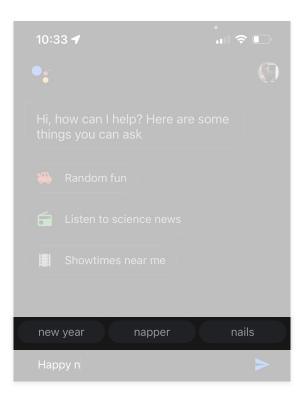
Continuous Monitoring with Federated ML



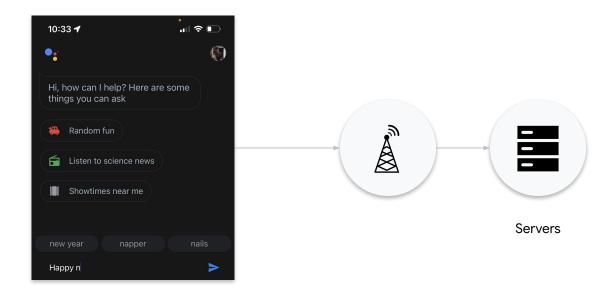
GBoard Example



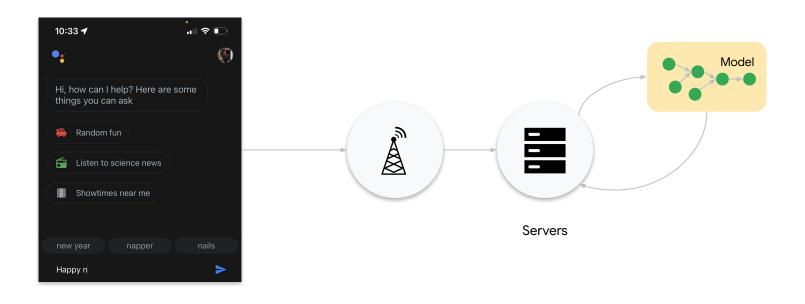
GBoard Example



How do we realize GBoard with MLaaS



How do we realize GBoard with MLaaS



How can **privacy be preserved**?

- Minimize
 - Avoid collecting unnecessary data, and dispose or delete data periodically
- Protect
 - Use encryption techniques to protect data
- Map the flow of information
 - Context, the type of information, and who has access
- Informed consent
 - Be transparent with users about how their data is being collected and used

Alternative approach?

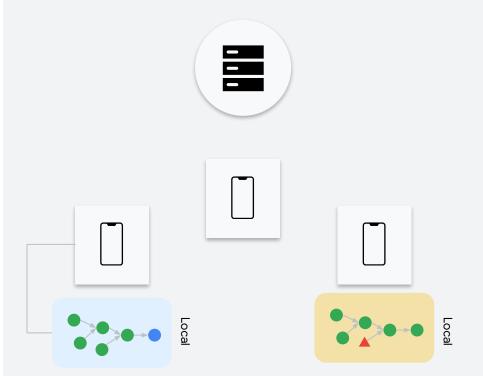
- The key flaw behind the prior approach to training is the sending of raw client data to central server
 - Server has access to raw client data, exposing clients to intrusion of privacy by central server



Central Servers

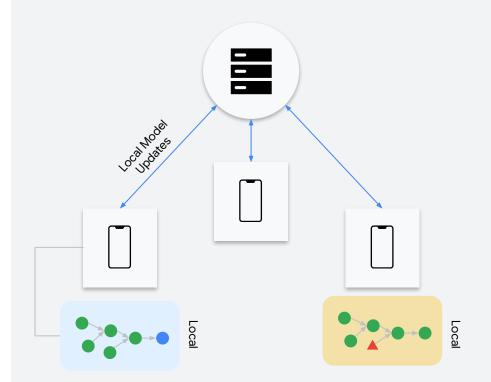
Federated ML

 Data is kept local to the endpoint device (data does not ever leave the device)

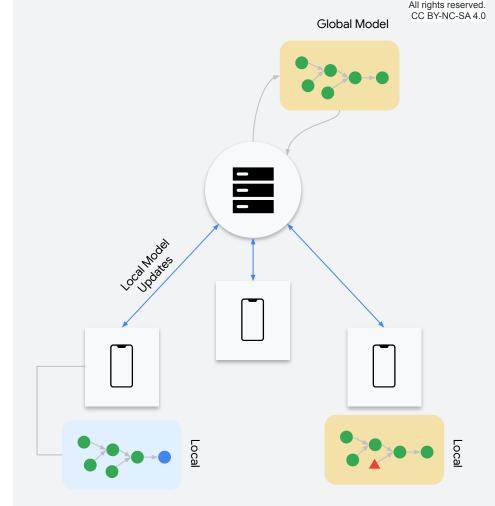


Federated ML

- Data is kept local to the endpoint device (data does not ever leave the device)
- Only local model updates are set to the central server



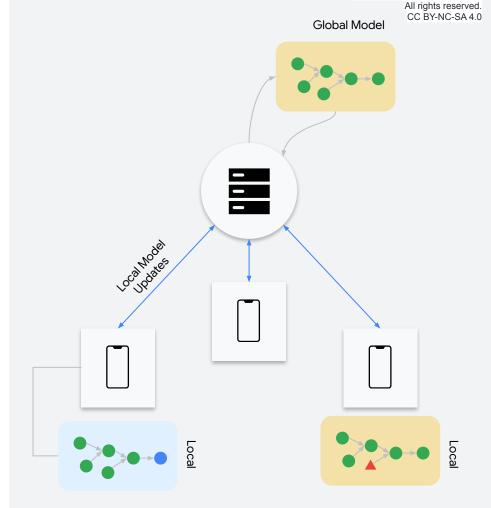
- Data is kept local to the endpoint device (data does not ever leave the device)
- Only local model updates are set to the central server
- Server creates a global model and sends it to the endpoints



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Why is Federated ML Useful?





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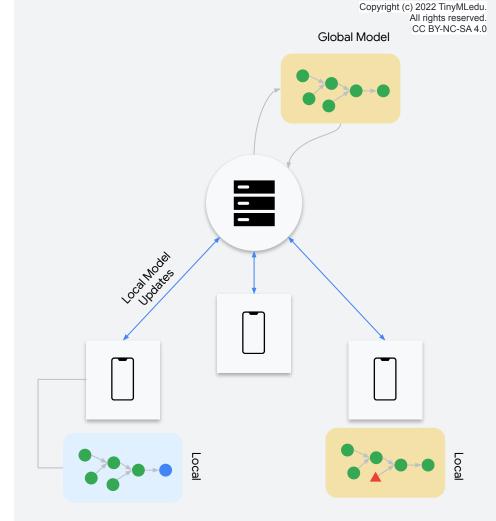
Why is Federated ML Useful?



Hyper-Personalized



Minimum Latencies



Why is Federated ML Useful?



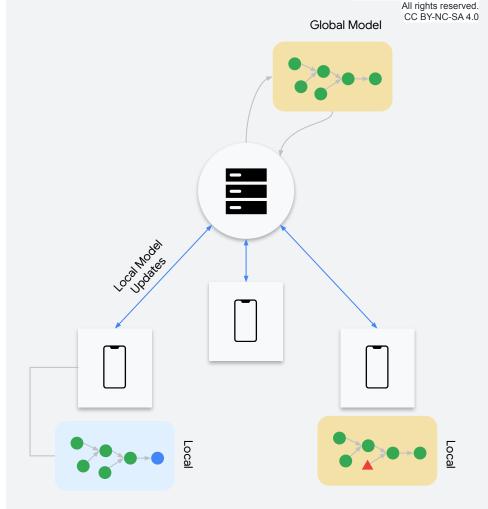
Hyper-Personalized



Low Cloud Infra Overheads



Minimum Latencies



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Hyper-Personalized



Low Cloud Infra Overheads

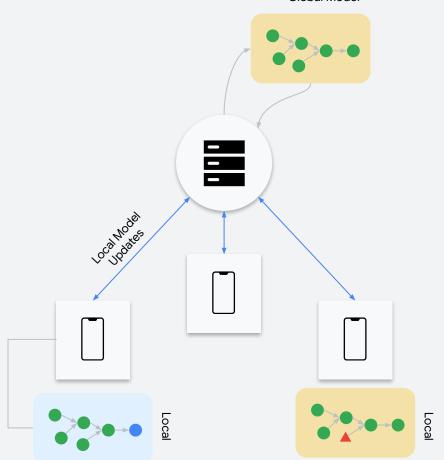


Minimum Latencies



Privacy Preserving

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Unbalanced training samples

Need a high # of endpoint devices/clients

Unbalanced training samples

Need a high # of endpoint devices/clients

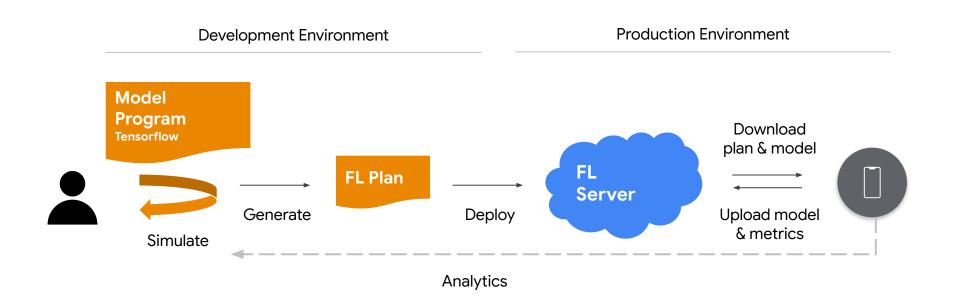
Unbalanced training samples

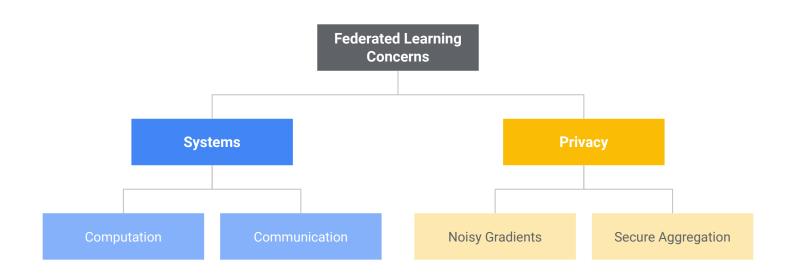
Need a high # of endpoint devices/clients

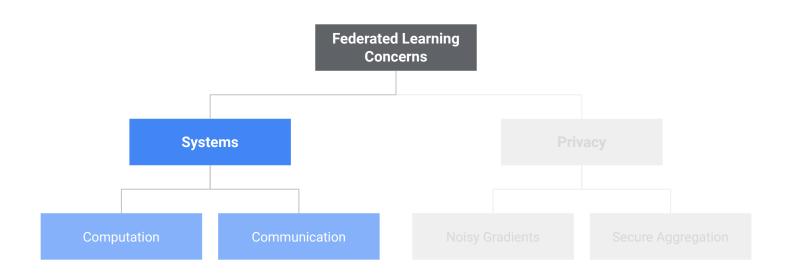
Unbalanced training samples

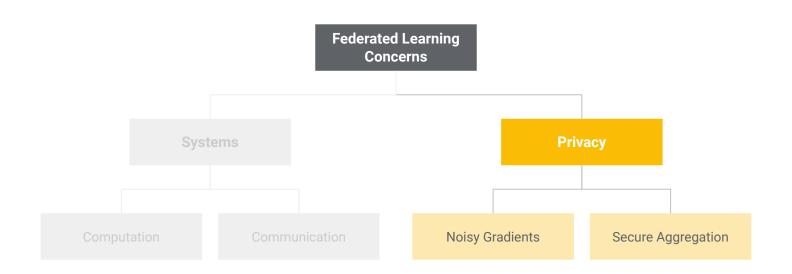
Need a high # of endpoint devices/clients

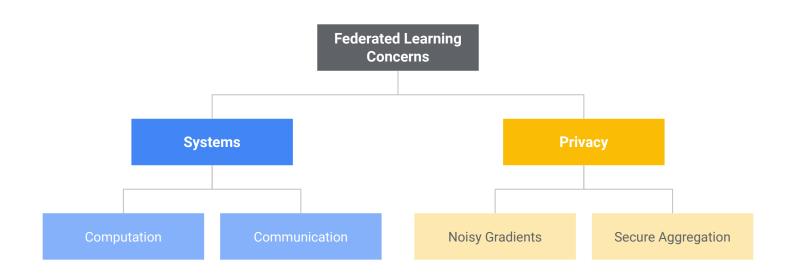
Federated Learning & MLOps











The MLOps **Personas**



ML Engineer



ML Researcher



Data Scientist



Data Engineer



Software Engineer



DevOps



Business Analyst

Send all of the raw clients' user
data to the server



Send all of the raw clients' user
data to the server

 All the ML model training is done in the remote cloud datacenters



 Key concern with sending raw data to the server: Privacy

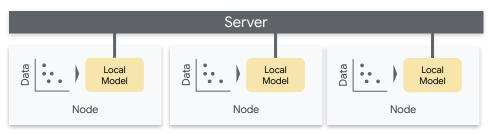


- Key concern with sending raw data to the server: Privacy
- Exposes user's raw data to the central server, which may potentially be compromised risking the loss of private data



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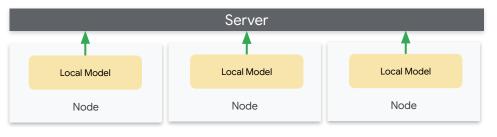
Receive model from server, start training.



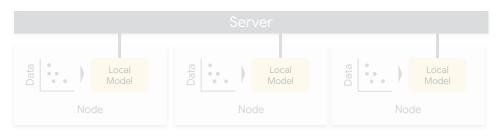
Receive model from server, start training.



Partially trained models \rightarrow server



Receive model from server, start training



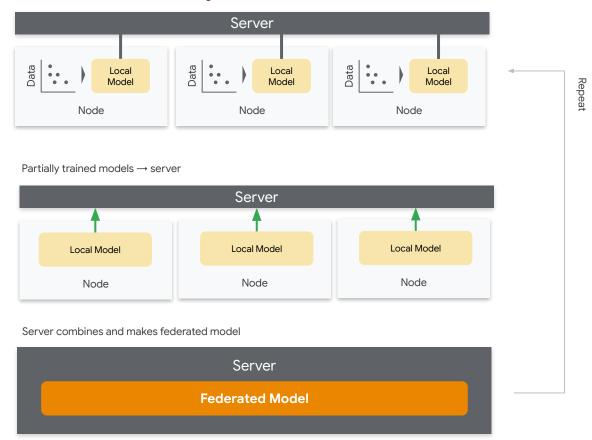
Partially trained models → serve



Server combines and makes federated model

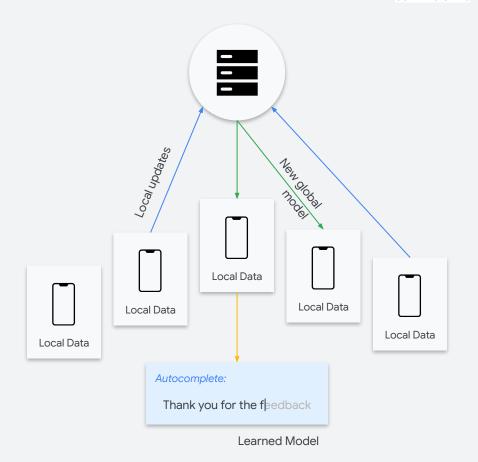


Receive model from server, start training.



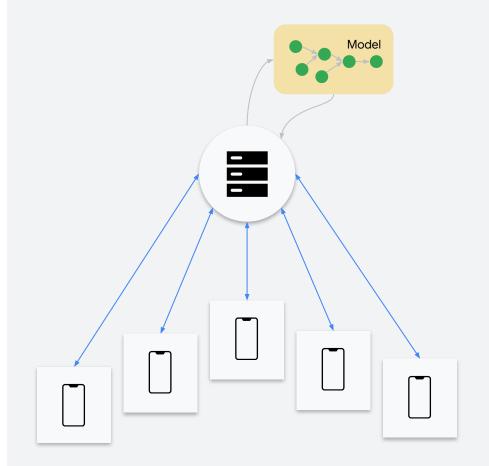
Federated Learning

 A method to collaboratively learn a shared model while keeping data on device



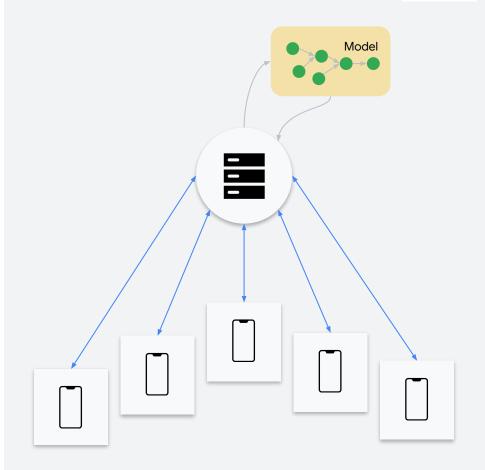
Traditional ML

 Data is aggregated from different sources at the server



Traditional ML

- Data is aggregated from different sources at the server
- Central server builds the machine learning model



Traditional ML

- Data is aggregated from different sources at the server
- Central server builds the machine learning model
- Central server distributes the global model to everyone

