# Let's Go, Private! Towards a Privacy-Preserving and Distributed Machine Learning System

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### Limitations

# Challenges

- Untrusted third-party infrastructures
- Common cryptographic schemes impose impractical overheads
- TEEs' performance decreases with the increase of computations, I/O operations and, the trusted computing base (TCB)

ML datasets and models are stored and processed in plaintext

Reducing the code base running inside enclaves

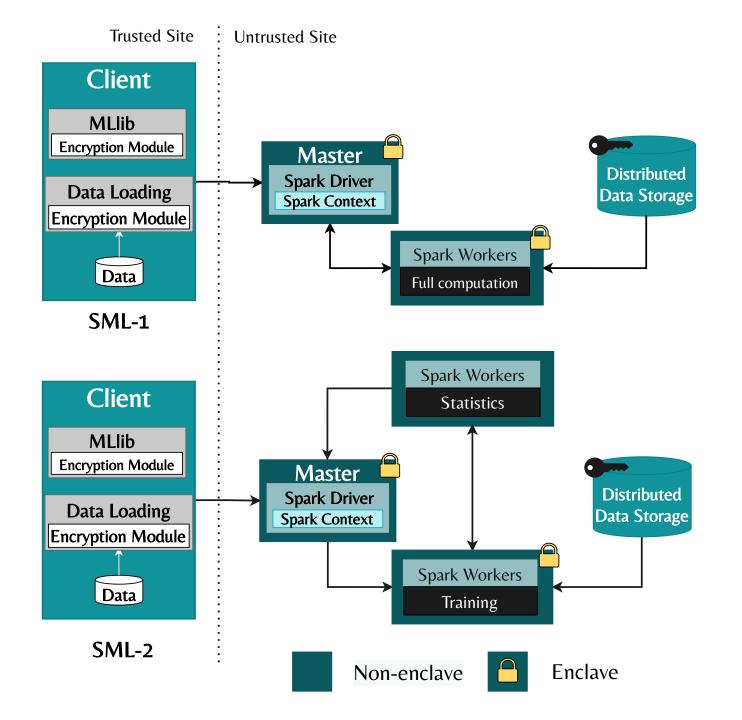
Reducing the number of operations to be performed inside the enclave

#### Goal

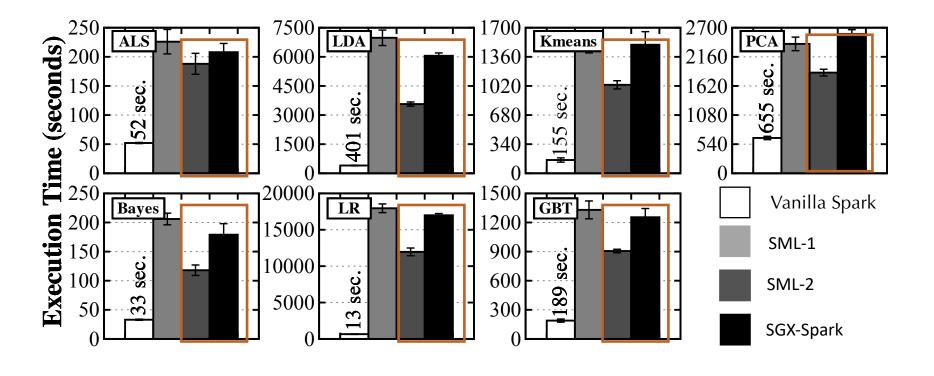
Design an end-to-end privacy-preserving and distributed machine learning framework

- Private large scale machine learning and data analysis
- Clients should trust third-party infrastructures while knowing that the computation performed over their data will **not reveal** any sensitive information

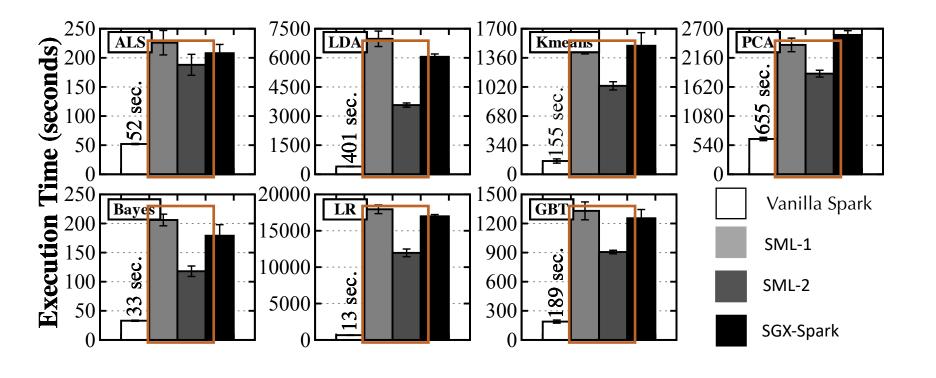
#### **Solution**



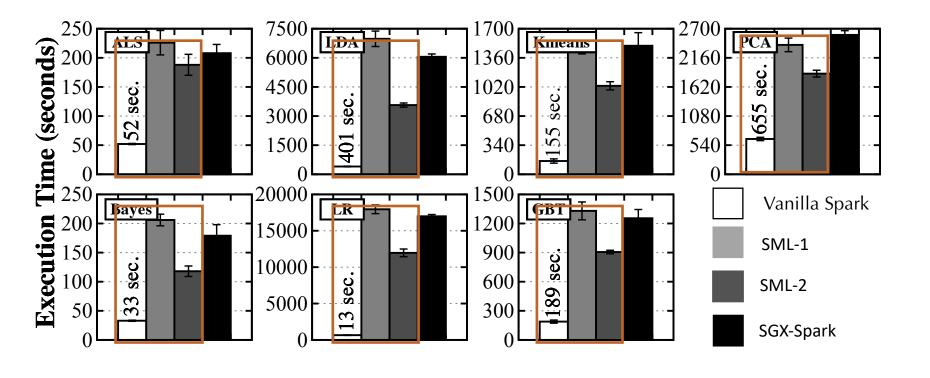
#### Results



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## **Next Steps**

- Focus on white-box access attacks
  - Increase security measures with focus on ORAM and Differential Privacy
- Real data use cases with a focus on genomic data