

隐私计算在大数据AI 领域的应用实践

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01隐私计算



隐私计算背景

个人的需求

• 隐私和安全的意识提高

隐私和安全合规要求

• 国外: 欧盟GDPR, 美国CCPA等

• 国内: 网络安全法, 数据安全法, 个人信息保护法等

宽松 隐私和安全的要求和管理



隐私计算背景

GDPR Fines Tracker & Statistics



总数:1087 (增长中)

总额:~110亿

单次最高:~50亿

前五罚款: Amazon, WhatsApp, Google,

Facebook



All data is from official government sources, such as official reports of national Data Protection Authorities.

https://www.privacyaffairs.com/gdpr-fines/



隐私计算现状

隐私计算成为热点

- 大量企业和投资涌入
- 大量研究成果
- 安全和隐私技术蓬勃发展
 - 差分隐私 (DP)
 - 可信执行环境 (TEE)
 - 同态加密 (HE)
 - 安全多方计算 (SMC)
 - 联邦学习 (FL/FML)

http://finance.people.com.cn/n1/2022/0424/c1004-32407072.html



兼顾安全与应用 2021年中国隐私计算市场规模4.9亿元

陈果静

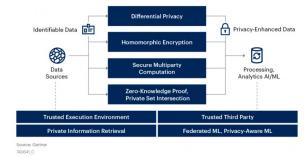
2022年04月24日09:15 | 来源: 经济日报

Tr 小字号

在数字经济时代,随着人工智能、大数据等产业的快速发展,数据要素的作用及其重要性愈发 凸显。与此同时,大量挖掘和收集用户数据与信息,金融领域用户数据信息泄露事件多发。

随着《数据保护法》《个人信息保护法》等的落地,金融业如何做到既保护数据安全,又充分 发挥数据资产价值、高效链接多方数据,成为一项亟需解决的课题。

Privacy-Enhancing Computation Techniques



Gartner.

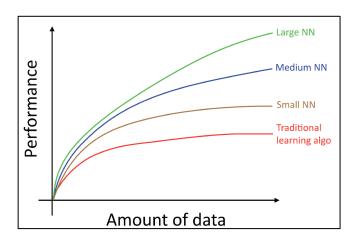






大数据AI背景

- 大数据框架和技术已经大规模普及
 - 易用性提高
 - 方向逐步细化
 - 存储、处理更多数据
 - 分析(查询)更多数据
 - 实时分析
 - 建模和预测(机器学习、深度学习)



"Machine Learning Yearning", Andrew Ng, 2016

- AI 无处不在
 - 从实验室走向生产环境
 - 应用于大规模、分布式大数据



获取/存储 分析/建模 部署/可视化 清洗/准备 集成的数据流水线 ووماوي Spark Spark Spark SQL Qlik Q Streaming HBASE RAPACHE Ka. A distributed streaming platform ANALYTICS Z KUDU **1** TensorFlow **⇔** RAY 数据分析 数据科学及人工智能



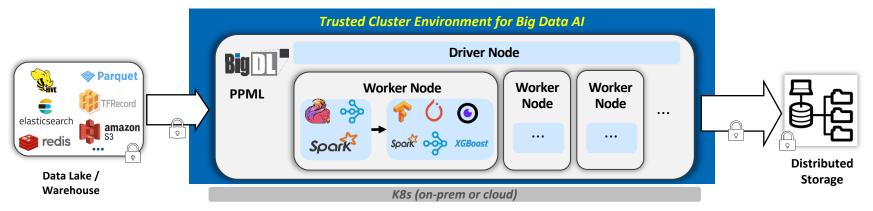
常见痛点:

- 能否兼容现有的应用
 - 现有的应用(数据分析和AI)能否直接迁移
 - 对其他应用和设施是否有冲击
- 能否处理大规模数据
 - 能否支持大规模数据
 - 计算效率是否足够好
- 能否解决数据孤岛问题



BigDL PPML: 可信的大数据AI

HW (SGX/TDX) Protected Secure Big Data AI, even on Untrusted Cloud

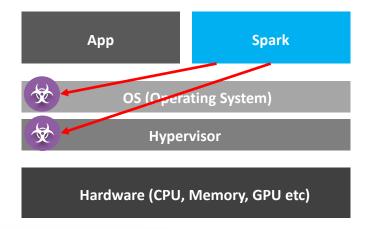


- Standard, distributed AI applications on encrypted data
- Hardware (Intel SGX/TDX) protected computation (and memory)
- End-to-end security enabled for the entire workflow
 - Provision and attestation of "trusted cluster environment" on K8s (of SGX nodes)
 - Secrete key management through KMS for distributed data decryption/encryption
 - Secure distributed compute and communication (via SGX, encryption, TLS, etc.)



Apache Spark中的安全

- 网络加密 (TLS/AES)
- 存储加密(AES)
- 计算(明文)



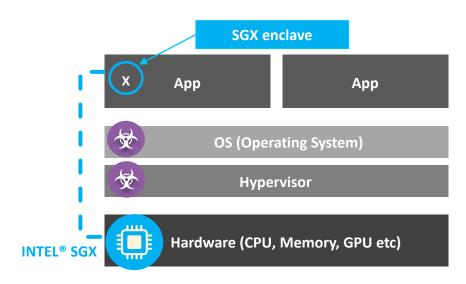
If OS/VM/Hypervisor/BIOS is hacked by adversaries, then they can dump sensitive data (input, temp, output etc) from Spark.



英特尔软件防护扩展SGX

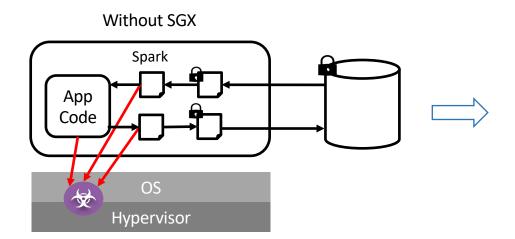
- 硬件级的可信执行环境(TEE)
- 相对小的攻击面
- 性能影响小
- 足够大的飞地(最大1TB)

已经被广泛测试、研究和部署





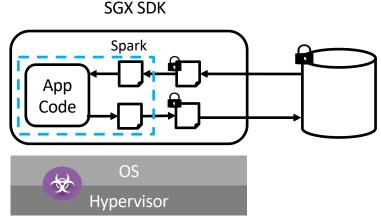
Secure Spark with SGX



攻击者可以获取到应用和敏感数据

缺点:

- 开发代价大
- 代码无法复用



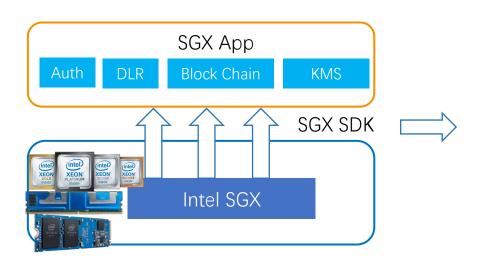
保护明文和敏感模块 攻击者无法获取明文数据

https://github.com/mc2-project/opaque-sql

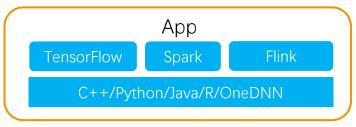


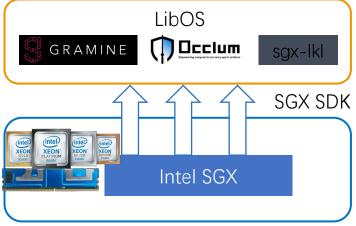


安全



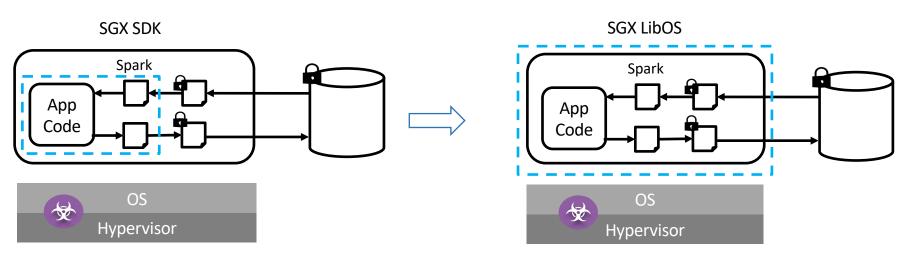
安全+易用性







Running unchanged Spark Applications in SGX



保护明文和敏感模块

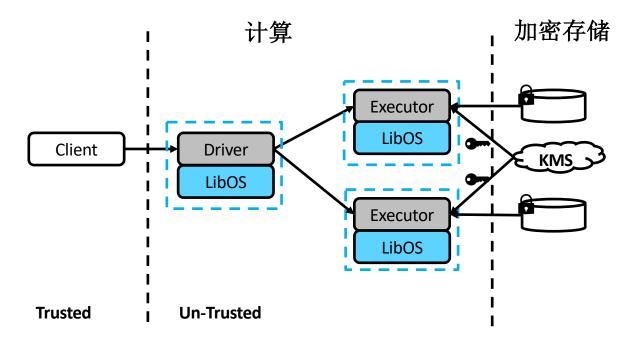
保护整个Spark

优点:不需要修改Spark和Spark应用





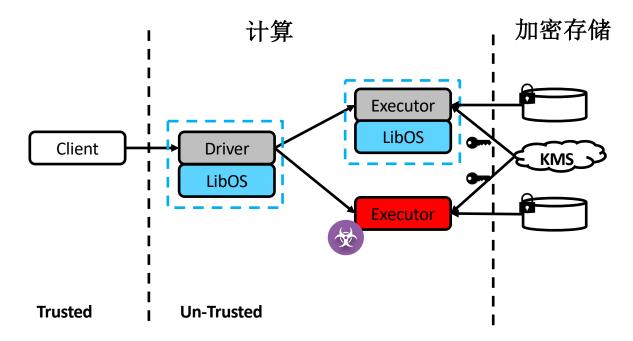
Running unchanged Spark Applications in SGX







Attack on distributed Spark

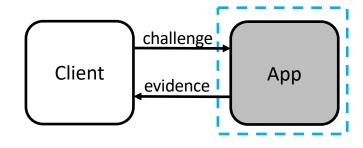






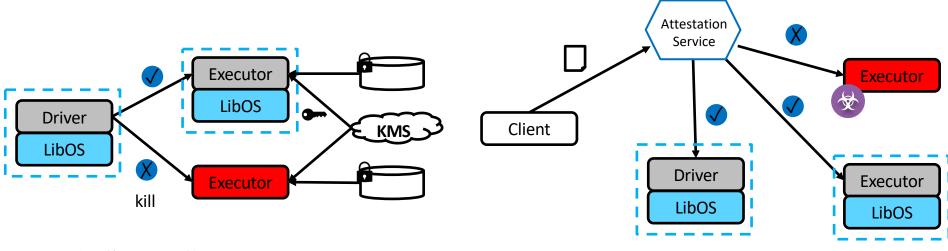
远程证明保证应用的完整性

- Attestation in short: <u>Verify if an application is running in SGX</u>
 - Application is expected
 - Within SGX
 - Running env is secured
 - ...



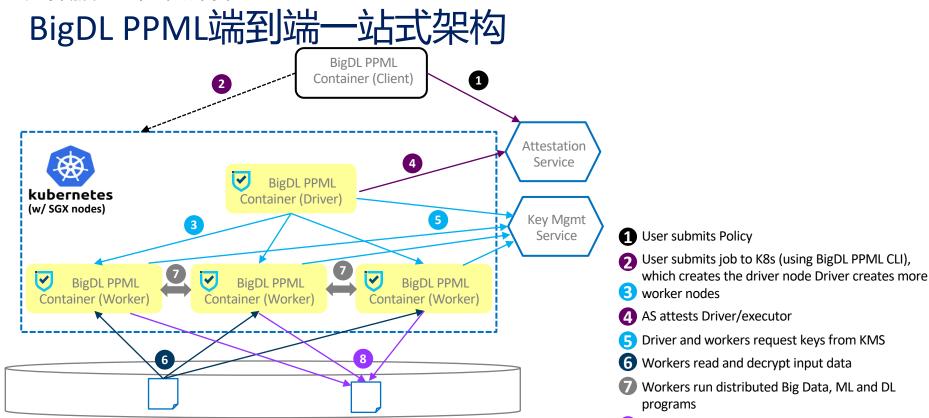
- Attestation result (verify evidence/quote)
 - Look good
 - Not good X

远程证明保证应用的完整性



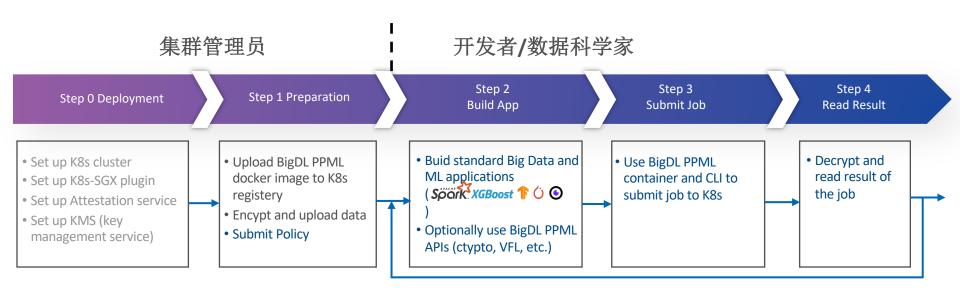
需要修改Spark的注册和Submit

无需修改Spark和Spark应用



8 Workers encrypt and write output data

BigDL PPML Workflow



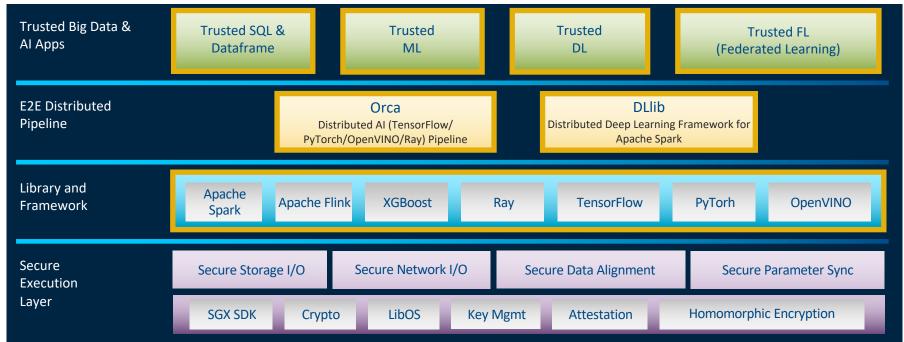
SGX相关的准备和开发

正常的建模和查询

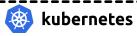
https://hub.docker.com/r/intelanalytics/bigdl-ppml-trusted-big-data-ml-scala-occlum

BigDL 隐私保护的机器学习

Secure & Trusted Big Data and AI











03 应用实践



实时的流计算-天池大赛



https://tianchi.aliyun.com/competition/entrance/531925/introduction

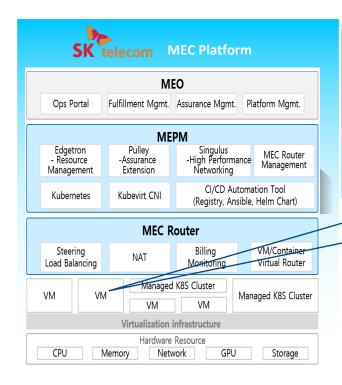


Alibaba, Intel and Occlum community co-host Kaggle-like PPML competition for spam detection in online e-commence recommendation.

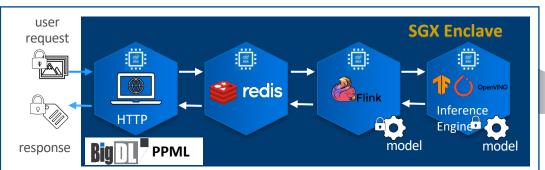




实时的流计算-SKT



https://networkbuilders.intel.com/solutionslibrary/referen ce-architecture-for-confidential-computing-on-skt-5g-mec



SKT Mobile Edge Computing provides common 5G services at the edge of the mobile telecommunication network. This POC runs Trusted Model Serving on BigDL PPML, providing secure, real-time, distributed DL model inference service across a cluster of Ice Lake servers

1300 image/sec

Secure Inference per MEC VM

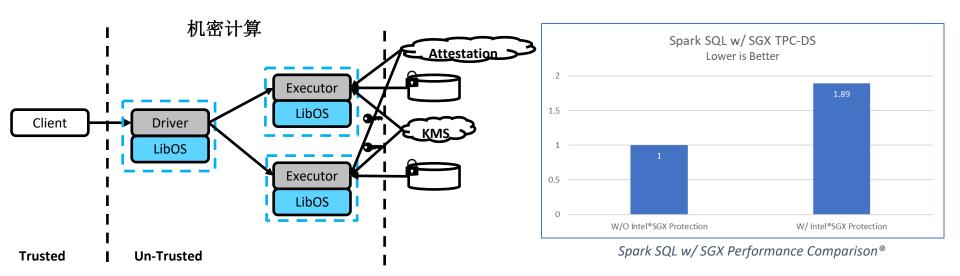
<5% overheads

F2F Inference Pipeline Overhead



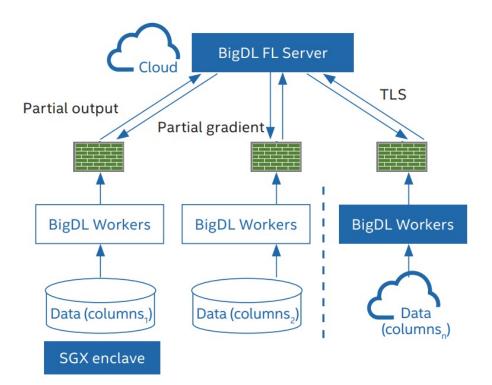


大规模数据分析SparkSQL TPC-DS





联邦学习



Trusted Federated Learning

- Build united model across different parities
 - Training data remain local
 - Aggregation temp/partial results
- Secured computation environment with SGX

Win-Win for all parties

- End users
- Enterprises
- Cloud Service providers

https://www.intel.cn/content/www/cn/zh/now/data-centric/sgx-bigdl-financial-big-data.html





04 总结和展望



总结和展望

隐私计算+大数据AI

- 若干痛点
- 用SGX构建安全的执行环境
 - LibOS帮助应用无缝迁移
 - 保证性能影响最小
 - 能够支持大规模数据
 - 联邦学习解决数据孤岛

BigDL PPML构建一站式的隐私计算方案



总结和展望

TEE发展趋势

- 易用性
 - TDX/Realm/SEV-SNP, 机密容器
- 安全性: TEEOS, Micro kernel
- 拓展性
 - IO的支持
 - 加速器的支持: GPU/QAT/FPGA



非常感谢您的观看

intel. | DataFun.

