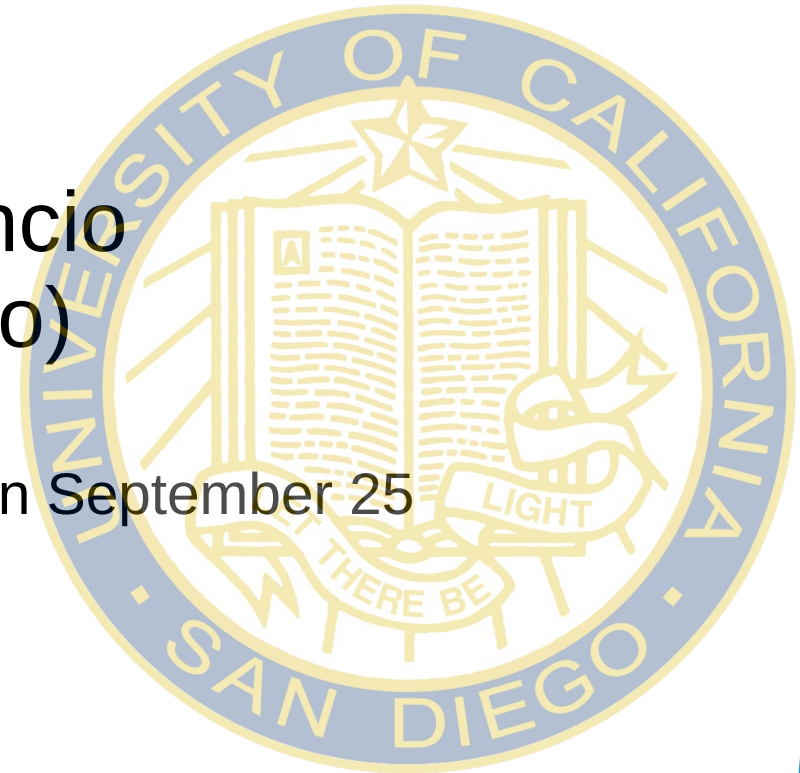


Overview of Fully Homomorphic Encryption: functionality and security models

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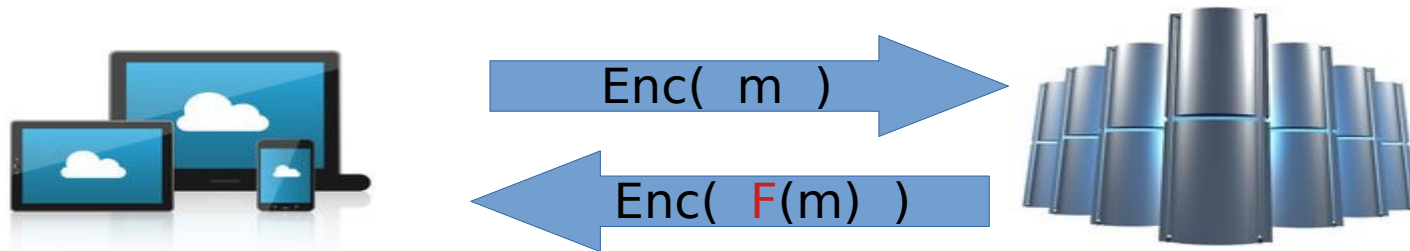


Fully Homomorphic Encryption

- Encryption: used to protect data at rest or in transit



- Fully Homomorphic Encryption: supports arbitrary computations (F) on encrypted data



FHE Timeline

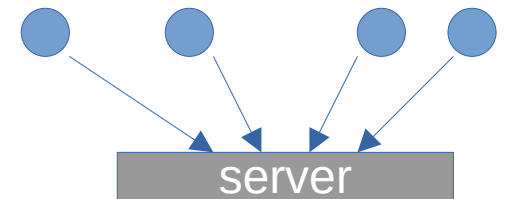
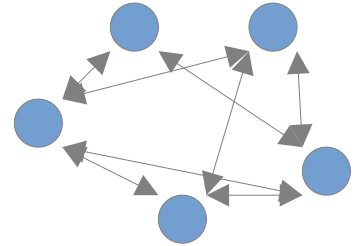
- 1978 – Rivest, Adleman, Dertouzos:
 - pose problem
- 2009 – Gentry:
 - **first** candidate solution
 - **bootstrapping** technique
- 2011 – Brakerski, Vaikuntanathan:
 - **first** solution based on standard lattice problems
- [BGV12,GHS12,GSW13,AP13/14,DM15,CGGI17,CKKS18,..., 2024]
 - new schemes, major efficiency improvements
 - Implementations: [SEAL, HElib, PALISADE, OpenFHE, HEAAN, Lol, FHEW, TFHE, LattiGo, ...]
 - all based on **lattices** and use **bootstrapping** technique

This talk

- Question: is FHE a good fit for a given application?
- Functionality
 - exact vs approximate computations
 - composability properties
- Security properties
 - passive vs active attacks
 - impact of decryption failures
- Advanced properties:
 - Verifiability, distributed decryption, etc.

FHE vs MPC

- Same problem: secure computation
- **MPC** (secure Multi Party Computation)
 - Data is “secret shared” among participants
 - Secure computation is done interactively
- **FHE** (Fully Homomorphic Encryption)
 - Data is protected using encryption scheme
 - Computation on encrypted data does not require interaction
 - Decryption key may be “secret shared”

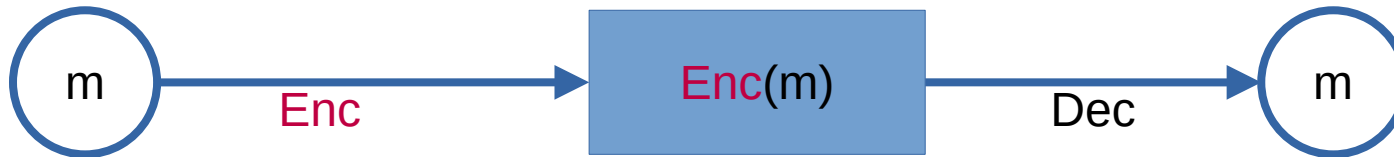


Use cases for FHE

- Public Key FHE scheme
- Workflow:
 - Multiple parties **encrypt their data locally**, under the same public key
 - Encrypted data is collected in encrypted form
 - **Computation** is performed **on encrypted data**
 - Final result is decrypted and shared with participants
- Examples:
 - Hospitals sharing patient data for joint medical study
 - Similarly for financial, or other sensitive data

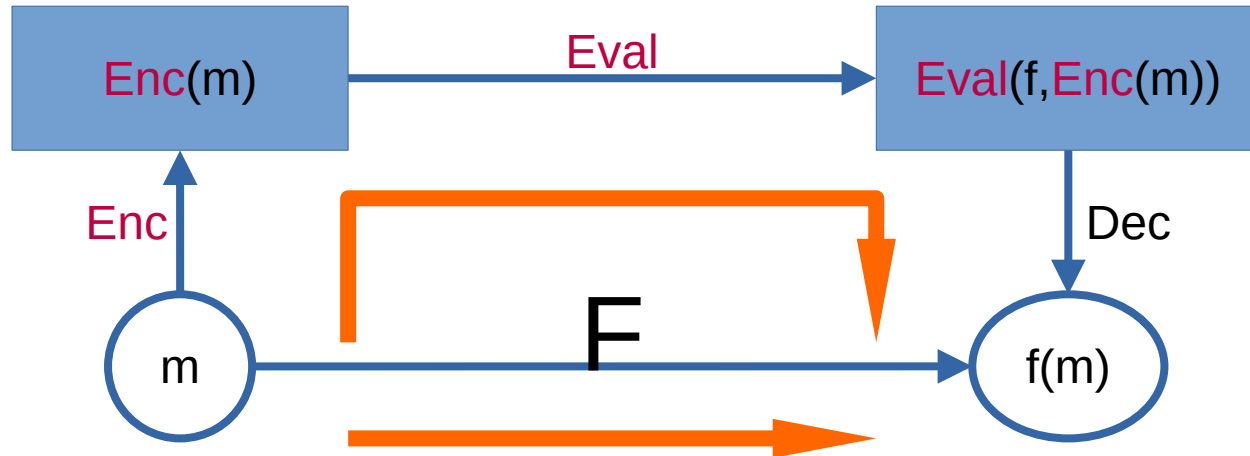
Encryption Scheme

- Syntax: (Gen, **Enc**, Dec)
- Correctness:
 - $(pk, sk) \leftarrow \text{Gen}$
 - $\text{Dec}_{sk}(\text{Enc}_{pk}(m)) = m$



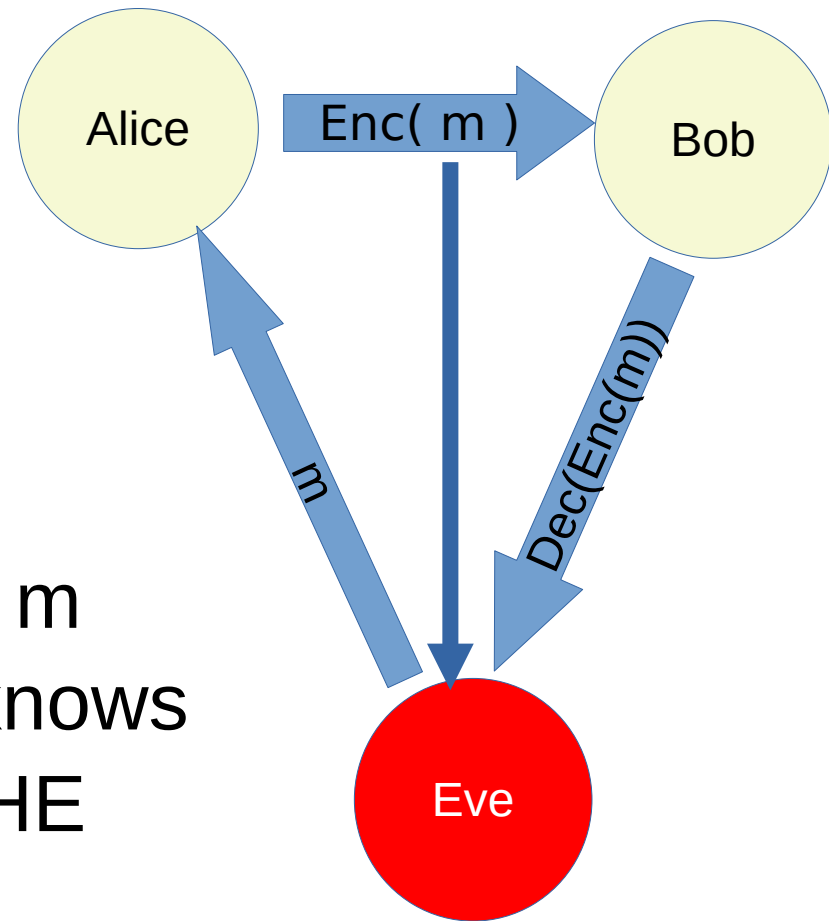
Fully Homomorphic Encryption

- FHE Scheme: (Gen, **Enc**, Dec, **Eval**)
 - $(pk, sk) \leftarrow \text{Gen}$
 - $\text{Dec}_{sk}(\text{Eval}_{pk}(F, \text{Enc}_{pk}(m))) = F(m)$



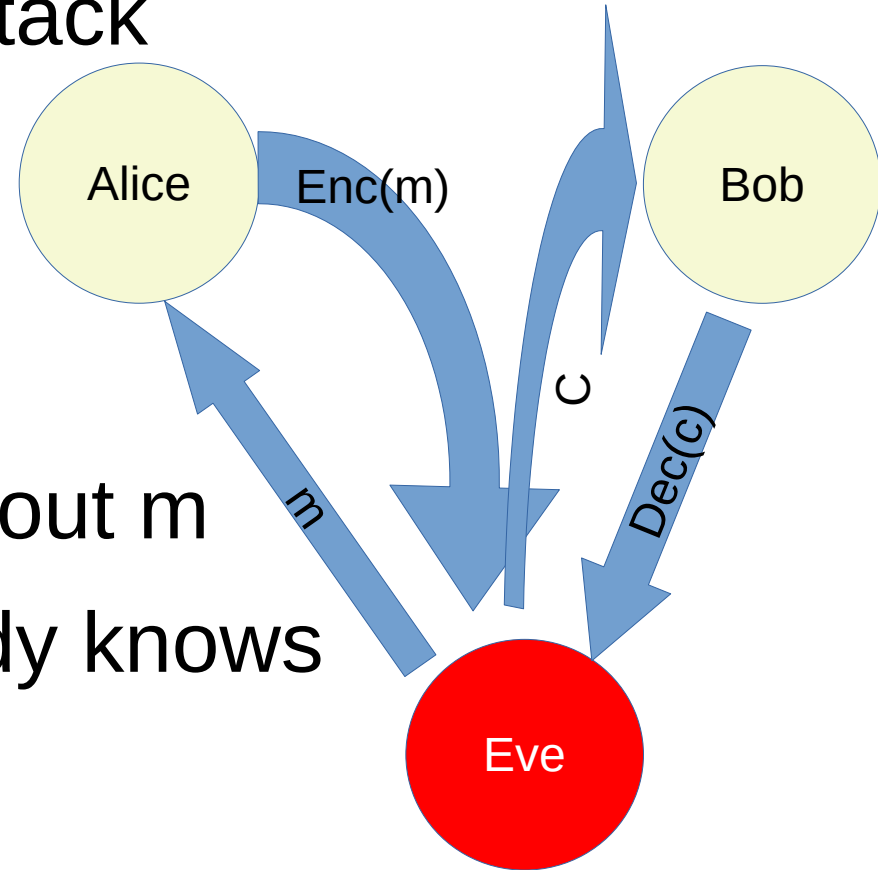
Passive attack model

- CPA: Chosen Plaintext Attack
- Adversary (eve) can:
 - Choose/influence message m
 - See the encryption $\text{Enc}(m)$
 - See result of decryption
 $\text{Dec}(\text{Enc}(m))=m$
- Still, cannot tell anything about m other than what she already knows
- Security definition applies to FHE



Active attack model

- CCA: Chosen Ciphertext Attack
- Adversary (eve) can:
 - See $\text{Enc}(m)$ of any m
 - See $\text{Dec}(c)$ of any c
- Still, cannot tell anything about m other than what she already knows



CPA/CCA security in Practice

- Remarks
 - Most applications require Active security
 - Active security implies Passive security
 - Active security can be achieved at reasonable cost (e.g., Fujisaki-Okamoto transform)
 - Standards (NIST, etc.) require Active security
 - All this is for regular (non-homomorphic) encryption
- What about Homomorphic Encryption?

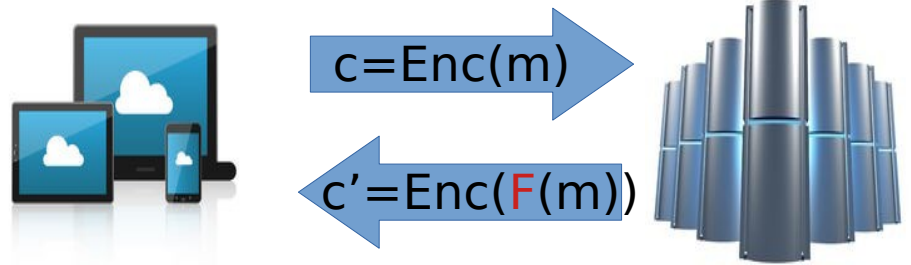
CCA security vs Non-Malleability

- CCA (active) security equivalent to non-malleability
 - Given $c = \text{Enc}(m)$, adversary **cannot compute** encryption c' of related message $\text{Dec}(c') = F(m)$
 - Intuition: If adversary cannot change c into c' , then active attack reduces to passive attack
- But this is exactly the opposite of FHE:
 - ability to change $\text{Enc}(m) \rightarrow \text{Enc}(F(m))$ is a useful feature!
 - FHE is perfectly malleable, and cannot be CCA secure

Concrete scenario

- Application:

- Store $c = \text{Enc}(m)$ on server
- Server computes $c' = \text{Eval}(F, c)$
- User decrypts final result $\text{Dec}(c') = F(m)$



- Questions:

- How do you know F was applied on correct c ?
- How do you know the server evaluated the correct F ?

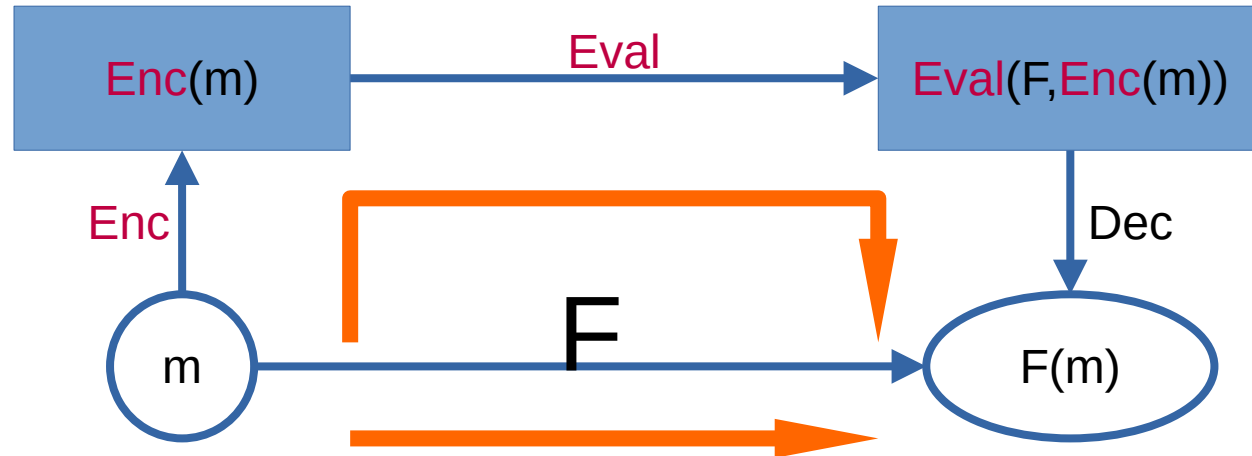
- Problem: **verifiable FHE**

- Can be addressed using zero-knowledge proofs, etc.
- **Active research area**, but not as mature as basic FHE

- Rest of this talk: focus on passive security

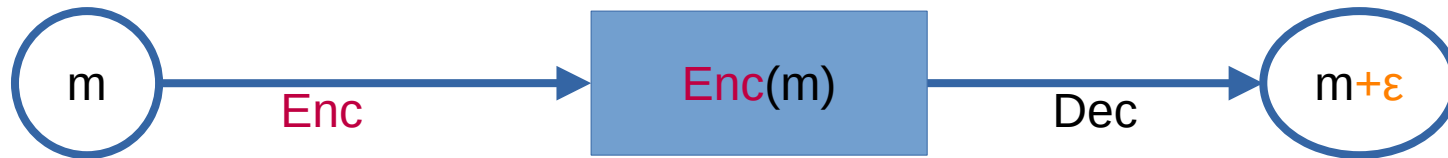
Fully Homomorphic Encryption

- FHE Scheme: $(\text{Gen}, \text{Enc}, \text{Dec}, \text{Eval})$
 - $(pk, sk) \leftarrow \text{Gen}$
 - $\text{Dec}_{sk}(\text{Eval}_{pk}(F, \text{Enc}_{pk}(m))) = F(m)$

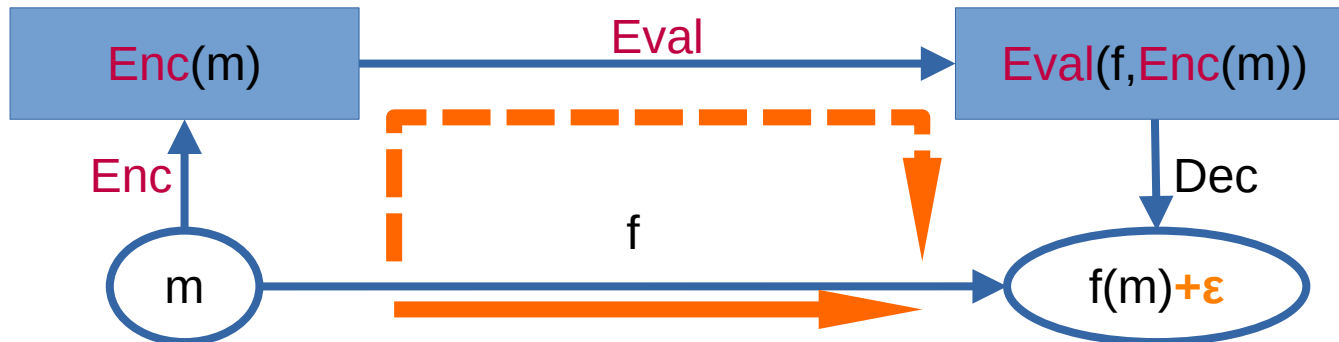


Approximate Encryption Scheme

- $\text{Dec}_{\text{sk}}(\text{Enc}_{\text{pk}}(m)) = m + \varepsilon$



- $\text{Dec}_{\text{sk}}(\text{Eval}_{\text{pk}}(f, \text{Enc}_{\text{pk}}(m))) = f(m) + \varepsilon$



Why approximate FHE?

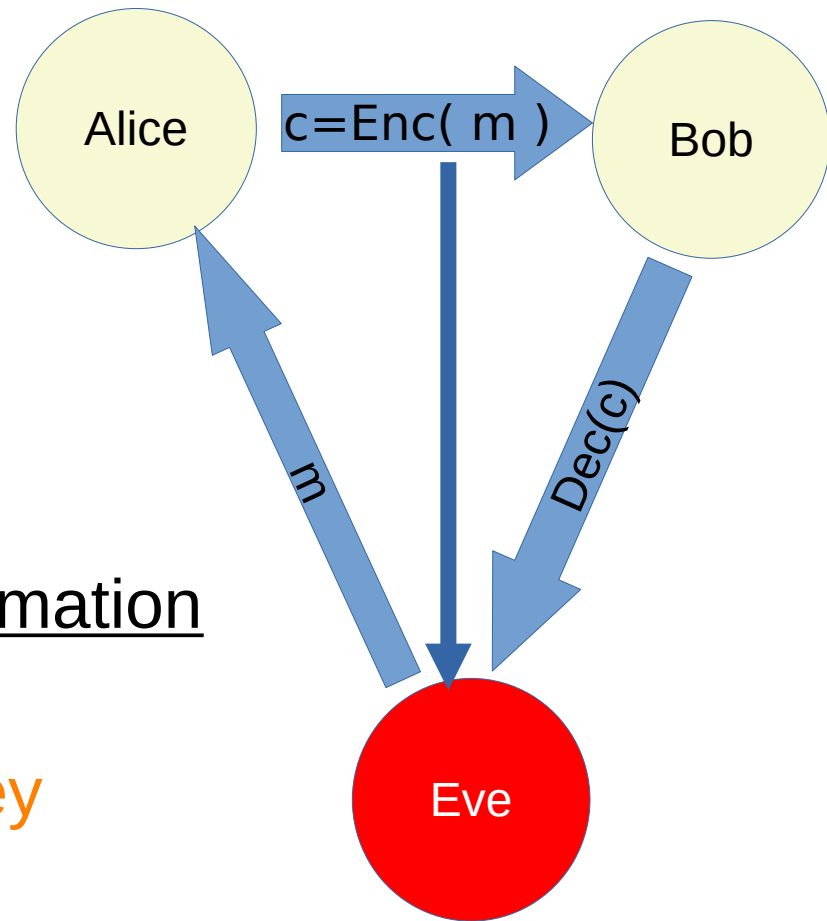
- Lattice cryptography (underlying FHE) is noisy:
 - In its most basic form $\text{Dec}'(\text{Enc}'(m)) = m + \epsilon$
 - Solution: use an error correcting code
 - $\text{Enc}(m) = \text{Enc}'(\text{encode}(m))$
 - $\text{Dec}(c) = \text{decode}(\text{Dec}'(c))$
 - $\text{Dec}(\text{Enc}(m)) = \text{decode}(\text{encode}(m) + \epsilon) = m$
- In FHE, homomorphic computations increase ϵ
 - Skipping encode/decode makes FHE much faster
 - In many applications, approximate results are acceptable (e.g., machine learning, statistics, etc.)

Approximate FHE

- [\[CKKS17\]](#): Homomorphic Encryption for Arithmetics on Approximate Numbers
 - Much more efficient than exact FHE
 - Satisfies standard CPA security definition
- Widely implemented and applied to machine learning, genome analysis, etc.
- [\[LM21\]](#): CKKS insecure under passive attacks!

Passive attack model

- CPA: Chosen Plaintext Attack
- Adversary (eve) can:
 - Choose/influence message m
 - See the encryption $c = \text{Enc}(m)$
 - See result of decryption $\text{Dec}(c)$
- For **exact** schemes
 - $\text{Dec}(c) = m$ gives no useful information
- For **approximate** schemes
 - $\text{Dec}(c) = m + \epsilon$ may **leak secret key**



Securing Approximate FHE

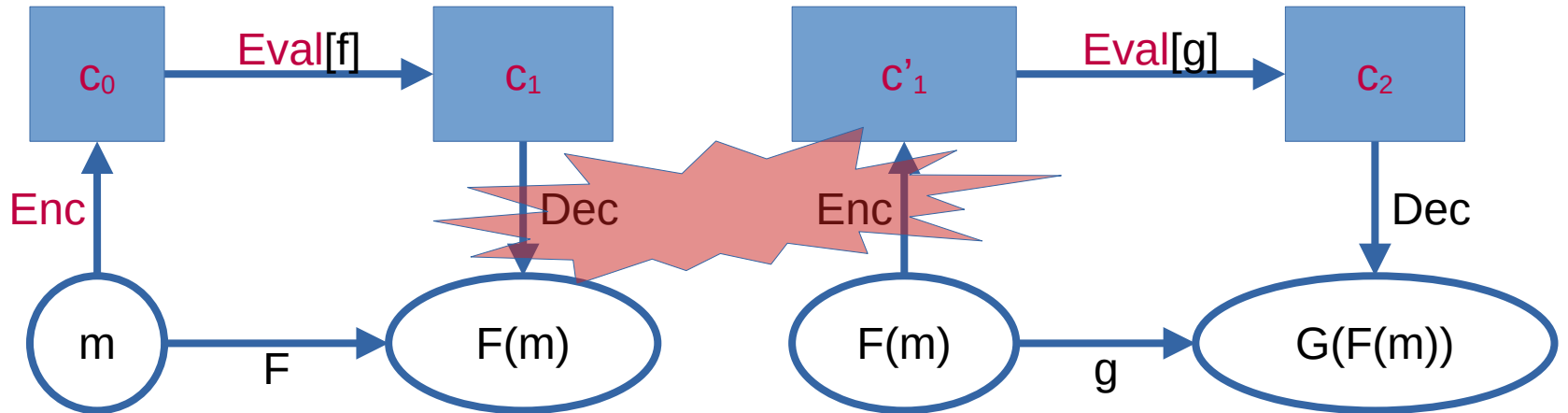
- [LM21]: new CPA-D security definition
 - Equivalent to CPA for exact schemes
 - Captures passive attacks when $\text{Dec}(c) = m + \epsilon$
- [LMSS22]:
 - Add extra noise to decryption $\text{Dec}(c) = \text{Dec}'(c) + \epsilon'$
 - Calibrate $\epsilon' \gg \epsilon$ to achieve CPA-D security
 - Reasonable cost, still more efficient than exact FHE

Composability

- $c_0 = \text{Enc}(m)$
- $c_1 = \text{Eval}(F, c_0)$
- $c_2 = \text{Eval}(G, c_1)$

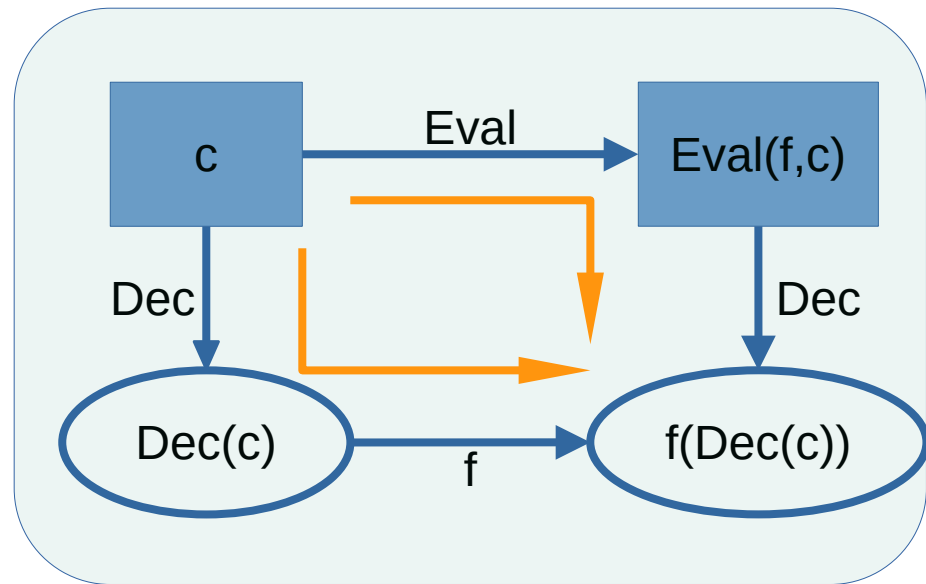
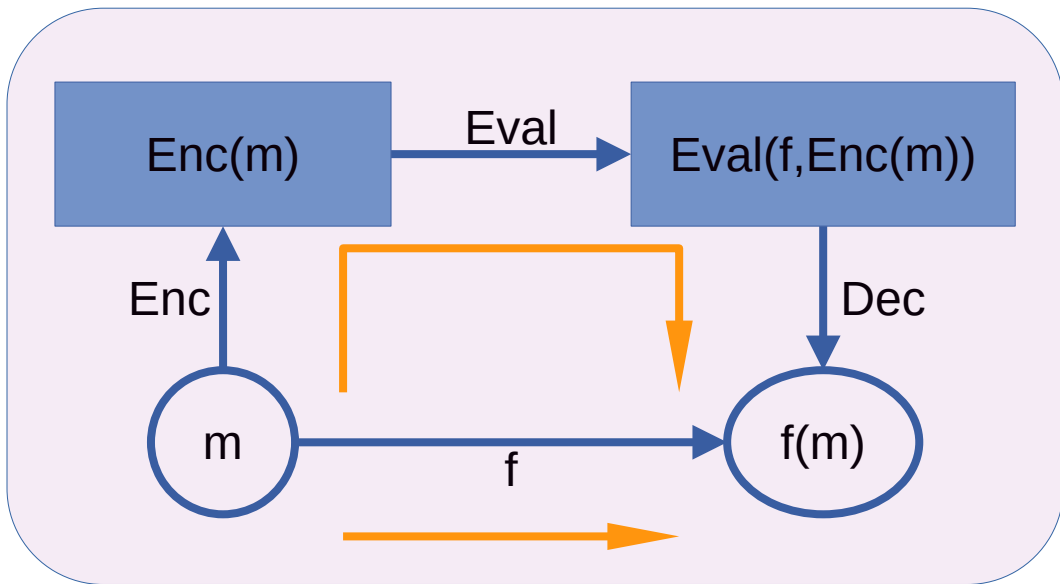
Question: Is $\text{Dec}(c_2) = g(f(m))$?

Answer: **not** necessarily



Standard vs Composable FHE

- Standard FHE: $\text{Dec}(\text{Eval}(F, \text{Enc}(m))) = F(m)$
- Composable FHE: $\text{Dec}(\text{Eval}(F, c)) = F(\text{Dec}(c))$



FHE Taxonomy

- Gentry: historical, but bootstrapping still relevant
- [BGV/BFV]:
 - Exact, operates on integer vectors (large message space)
 - Slow bootstrapping, but high bandwidth (SIMD)
- [DM/CGGI] (FHEW/TFHE):
 - Exact, fast, composable, single bit operations
 - Active research on SIMD extensions
- [CKKS]:
 - Approximate, operates on real vectors (large message space)
 - Faster than BGV/BFV at cost of approximate results
 - Requires LMSS noise padding to achieve security

Noise estimation / padding

- [LMSS]: securing CKKS requires adding noise
 - $\text{Dec}(c) = \text{Dec}'(c) + \epsilon'$
- How much noise?
 - Larger ϵ' gives more security
 - Smaller ϵ' gives more accurate results
 - $\epsilon' \gg \epsilon$: should be larger than c 's noise $\text{Dec}'(c) = m + \epsilon$
- Question: how big is ϵ ?

Estimating ciphertext noise

- $\text{Dec}(\text{Enc}(m)) = m + \varepsilon$, for small ε chosen by Enc
- $\text{Eval}(F, \text{Enc}(m)) = F(m) + \varepsilon$, for larger ε , dependent on f
- In (lattice-based) FHE:
 - Parameters (encode/decode) should be set large enough to correct ciphertext ε noise
 - Large ε has negative effect on efficiency
 - Even more so for Approximate FHE, which requires adding extra $\varepsilon' \gg \varepsilon$

Application-aware FHE [AAMP24]

- In many applications,
 - function F is fixed, and known in advance
 - E.g., common statistics: mean, average, standard deviation of encrypted data set
- Good trade-off between security and efficiency:
 - Use function F to estimate ciphertext noise ϵ
 - Generate FHE parameters specific to f, ϵ
- Warning: if $c' = \text{Eval}(F', c)$ is called with different F' :
 - $\text{Dec}(c')$ may be incorrect
 - $\text{Dec}(c')$ may leak information about secret key

Distributed FHE decryption

- FHE: $c = \text{Enc}(m) \rightarrow c' = \text{Enc}(F(m))$
 - both input and output are encrypted
 - Good and bad at the same time
- Secret (decryption) key sk :
 - Needed to recover final result $F(m) = \text{Dec}_{sk}(c')$
 - It also allows to decrypt original input $m = \text{Dec}_{sk}(c)$
 - Single point of failure
- Solution: secret share sk , and decrypt using MPC

Threshold FHE

- FHE with specialized distributed Dec protocol
 - Lattice-based encryption is “key homomorphic”
 - $\text{Dec}'(\mathbf{sk}_1 + \dots + \mathbf{sk}_n, c) = \text{Dec}'(\mathbf{sk}_1, c) + \dots + \text{Dec}'(\mathbf{sk}_n, c)$
- How to share/use secret key \mathbf{sk} :
 - Pick random $\mathbf{sk}_1 + \dots + \mathbf{sk}_n$ such that $\mathbf{sk}_1 + \dots + \mathbf{sk}_n = \mathbf{sk}$
 - Each share holder computes $\mathbf{d}_i = \text{Dec}'(\mathbf{sk}_i, c)$
 - Results are combined into $\text{decode}(\mathbf{d}_1 + \dots + \mathbf{d}_n) = m$
- Problem: \mathbf{d}_i are noisy and may leak information about \mathbf{sk}_i
- Solution, similar to approx. FHE:
 - Add noise $\text{Dec}(\mathbf{sk}_i, c) = \text{Dec}'(\mathbf{sk}_i, c) + \varepsilon_i$

Concluding Remarks

- Current FHE implementations:
 - promising technology, potentially useful in many critical applications
 - major efficiency gains during the last 15 years
 - reasonably efficient to be used in practice
- FHE is a technical tool, to be used with care
 - Current schemes target passive security
 - Even passive security can already be quite tricky for approximate/threshold schemes
 - Current FHE research is about much more than just efficiency improvements

Some References

- [BGV] <https://ia.cr/2011/277>
- [DM] <https://ia.cr/2014/816>
- [CKKS] <https://ia.cr/2016/421>
- [CGGI] <https://ia.cr/2018/421>
- [LM] <https://ia.cr/2020/1533>
- [LMSS] <https://ia.cr/2022/816>
- [AAMP] <https://ia.cr/2024/203>