

# Machine learning over encrypted data

**Rener Oliveira**  
**Rodrigo Targino**

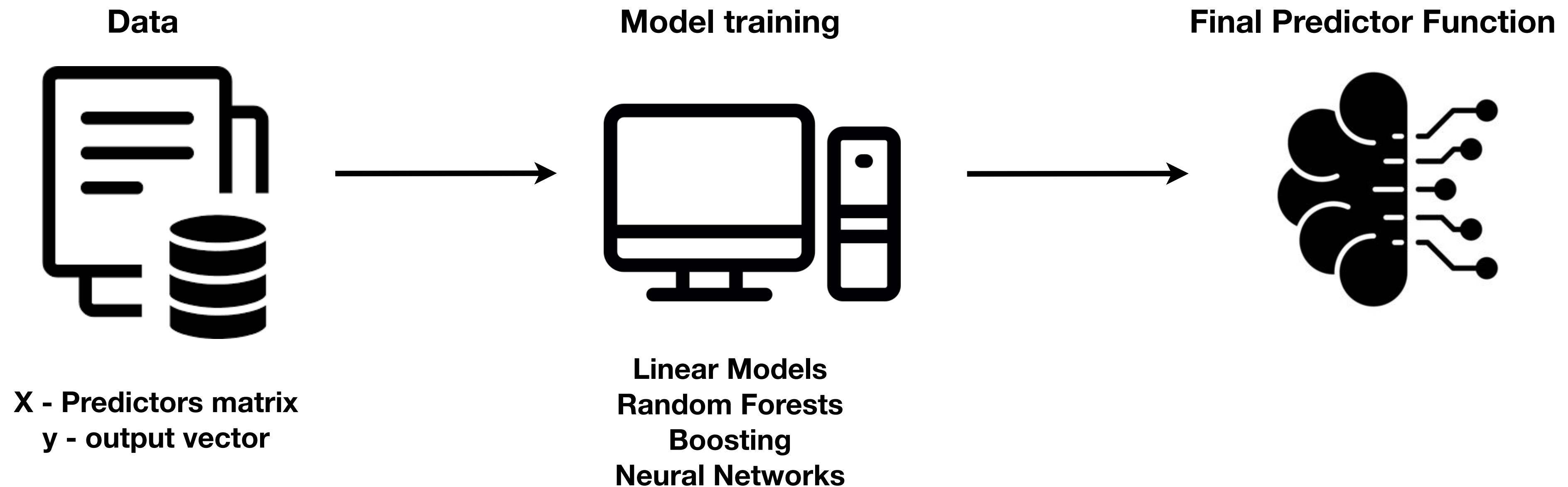
**June 27th, 2022**

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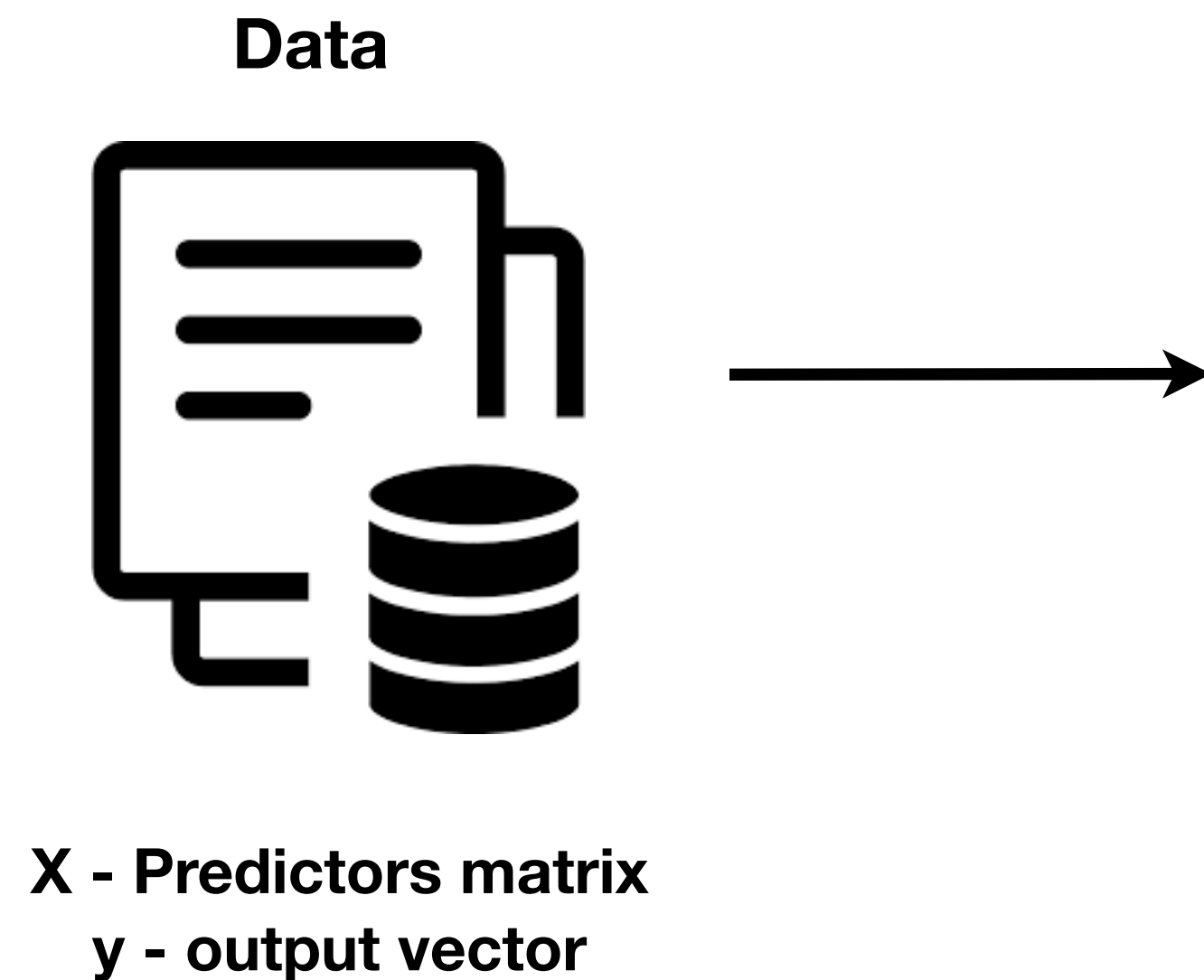
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- Medical records
- Financial data
- Genetic data

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`$&h%$8` - Encrypted target vector

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**Maybe :)**

# Privacy Homomorphisms, 1978

ON DATA BANKS AND PRIVACY HOMOMORPHISMS

*Ronald L. Rivest*

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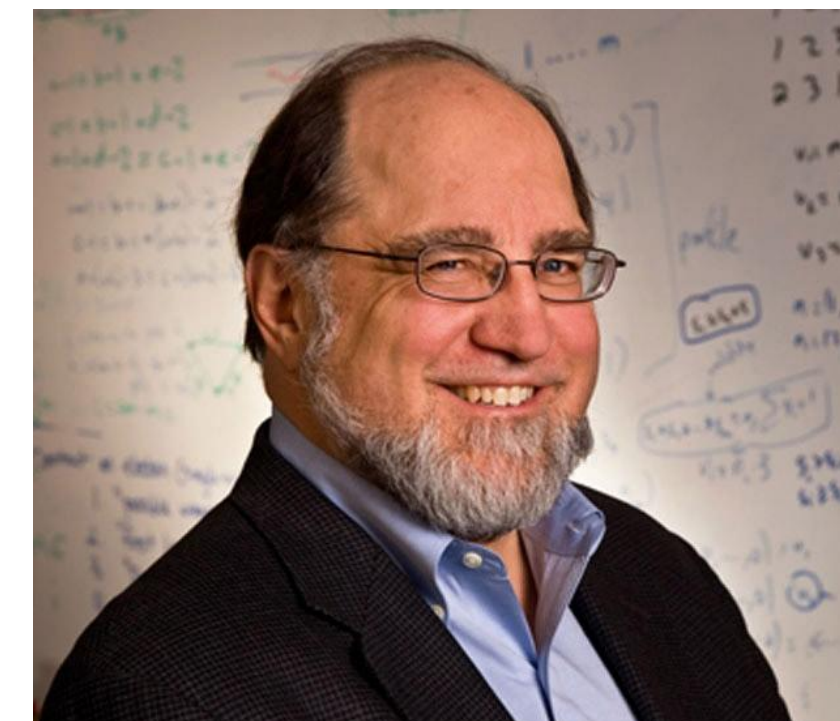
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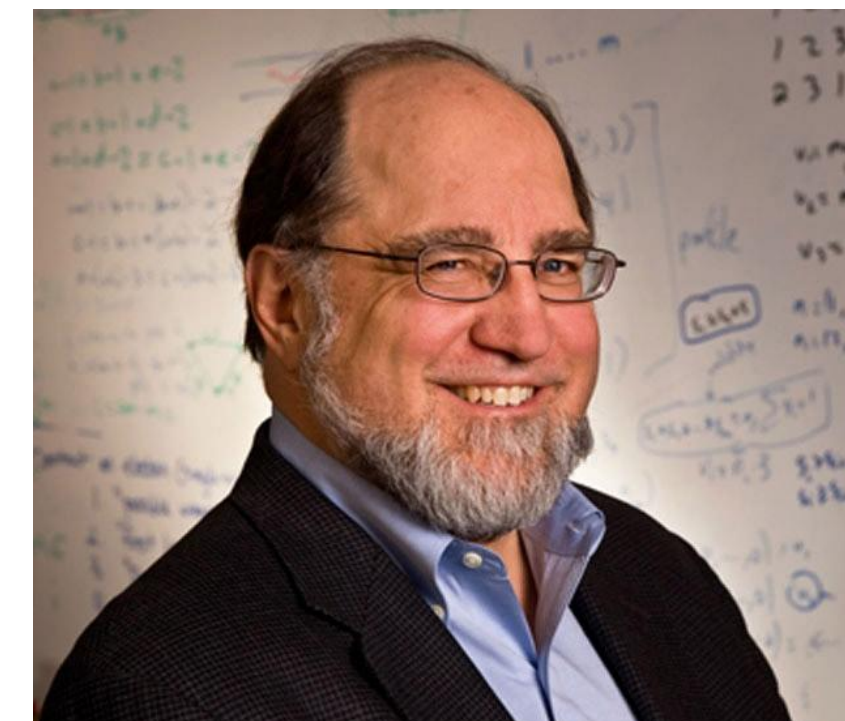
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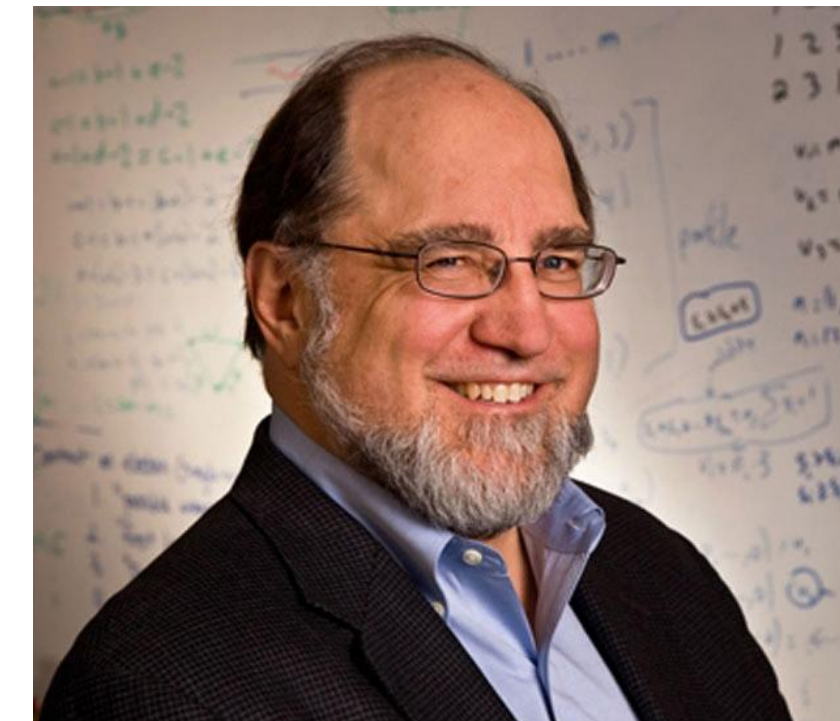
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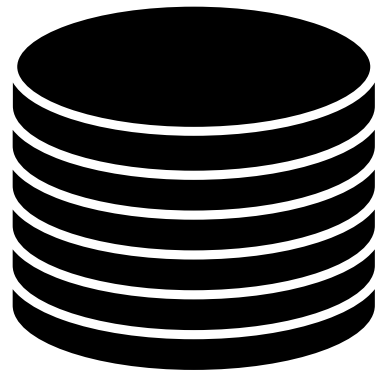


limitations on what can be accomplished, we shall see that it appears likely that there exist encryption functions which permit encrypted data to be operated on without preliminary decryption of the operands, for many sets of interesting operations. These special encryption functions we call "privacy homomorphisms"; they form an interesting subset of arbitrary encryption schemes

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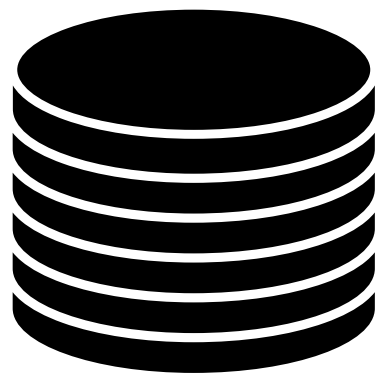
Message space:  
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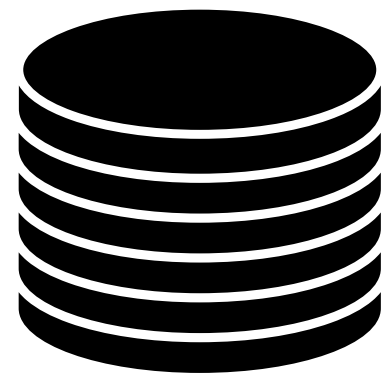
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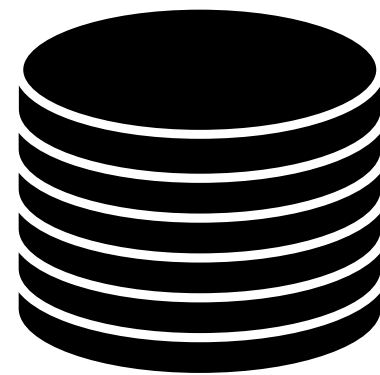


pk = public key

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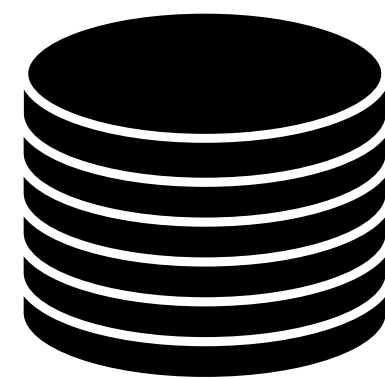


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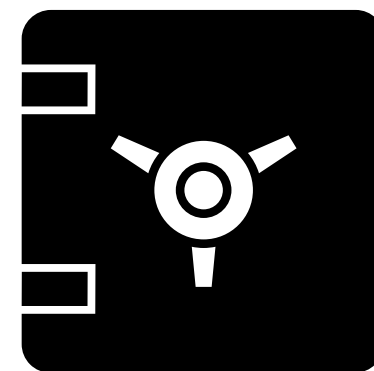
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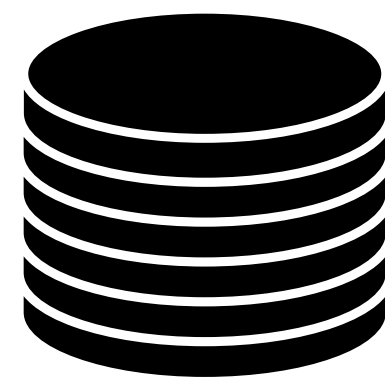
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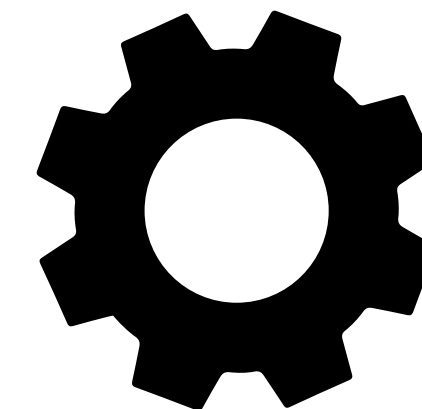
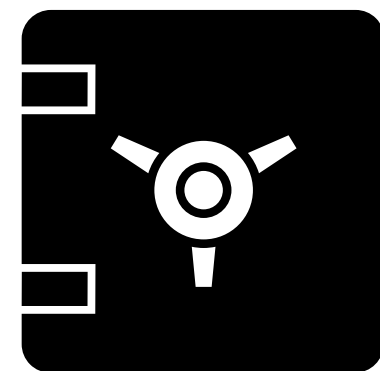
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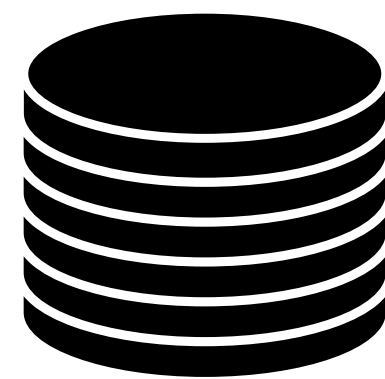
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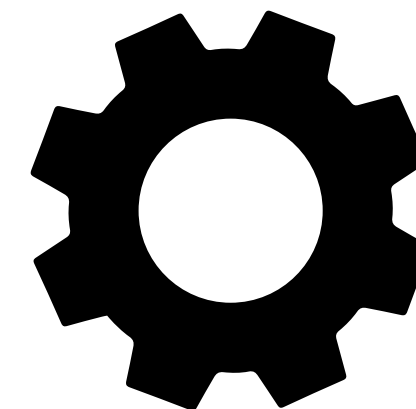
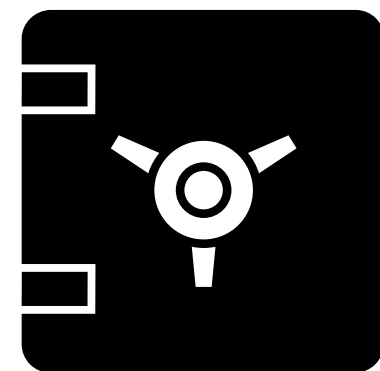
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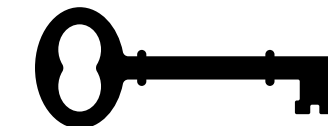
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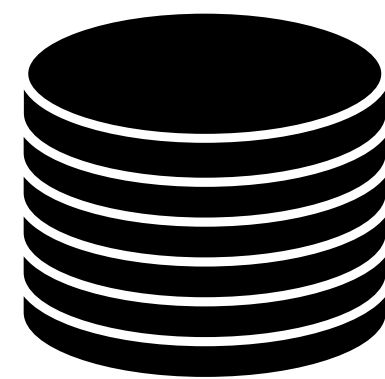
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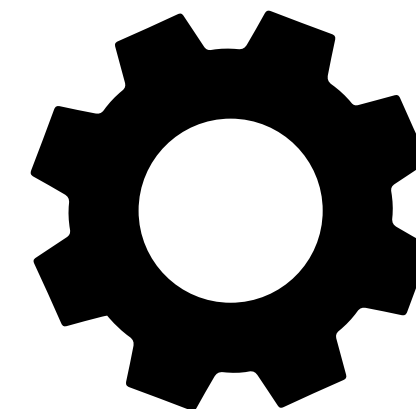
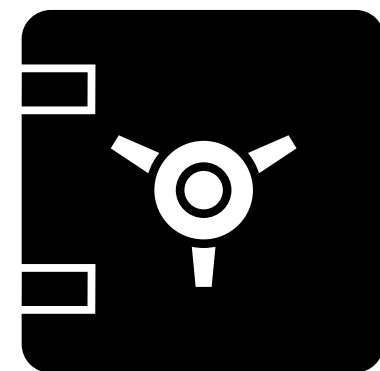
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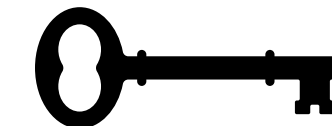


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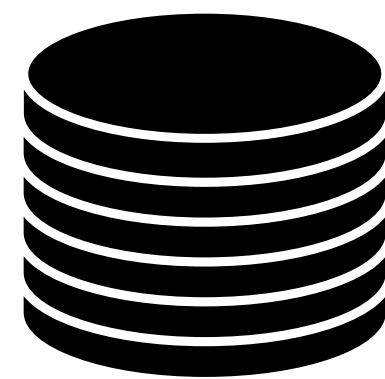
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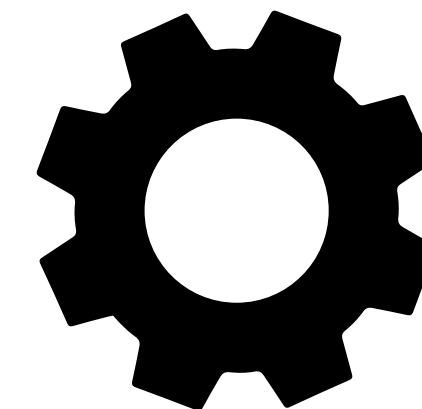
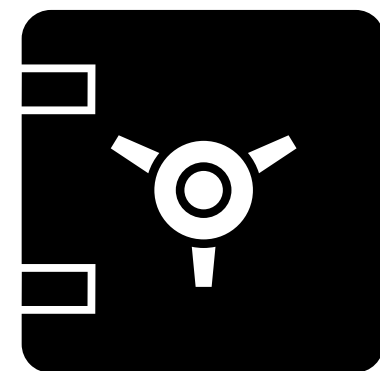
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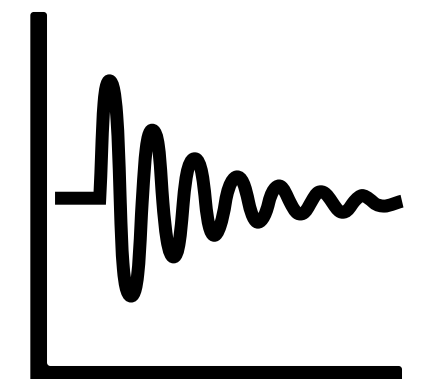
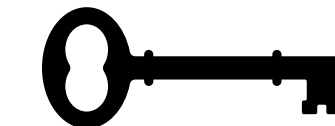


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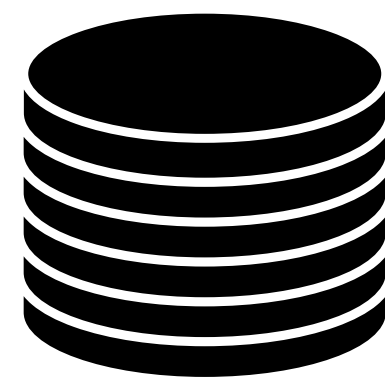
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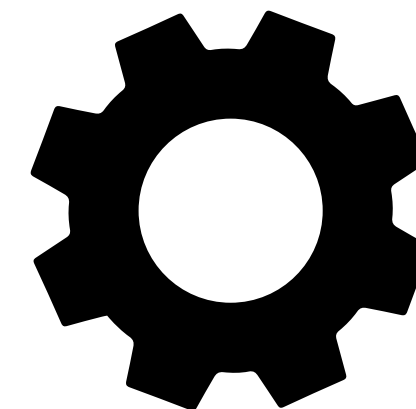
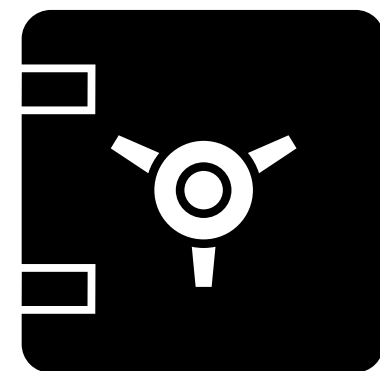
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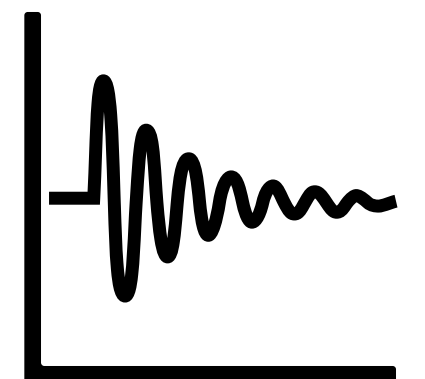
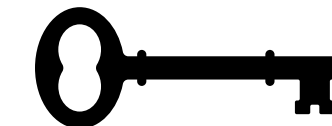


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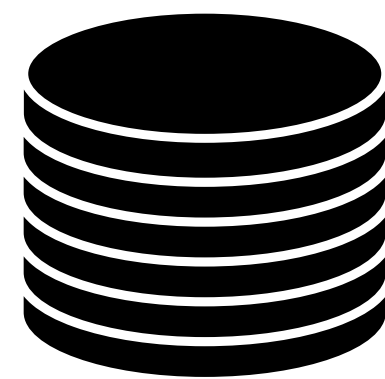
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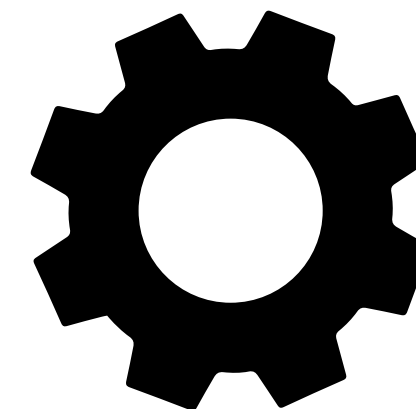
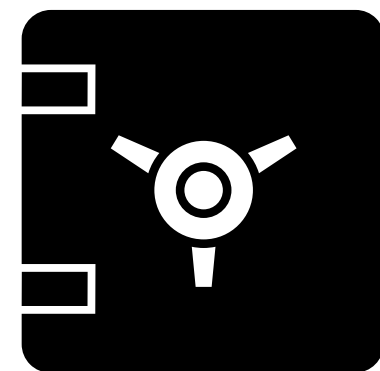
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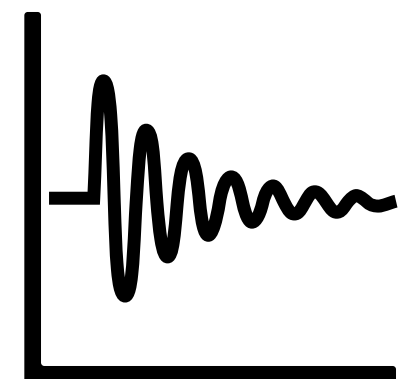
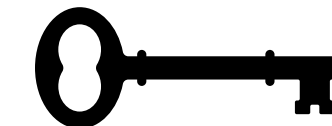


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# Fully Homomorphic Encryption

**Craig Gentry built the first FHE,  
on his Ph.D. Thesis in 2009**

**Message space:  $\{0,1\}$   
Permitted functions:  
additions and multiplications**





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- Encrypt the same message 2+ times must generate different cyphertexts
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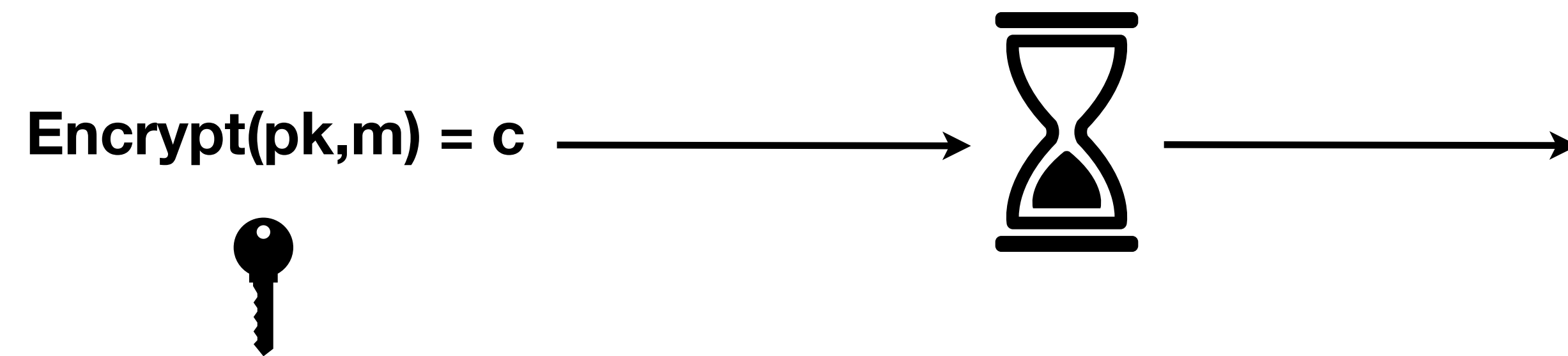
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$\text{Encrypt}(\text{pk}, m) = c$



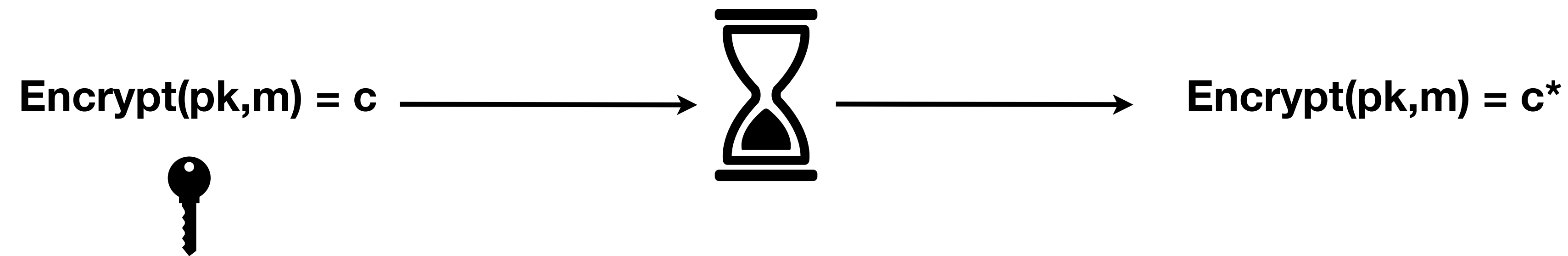
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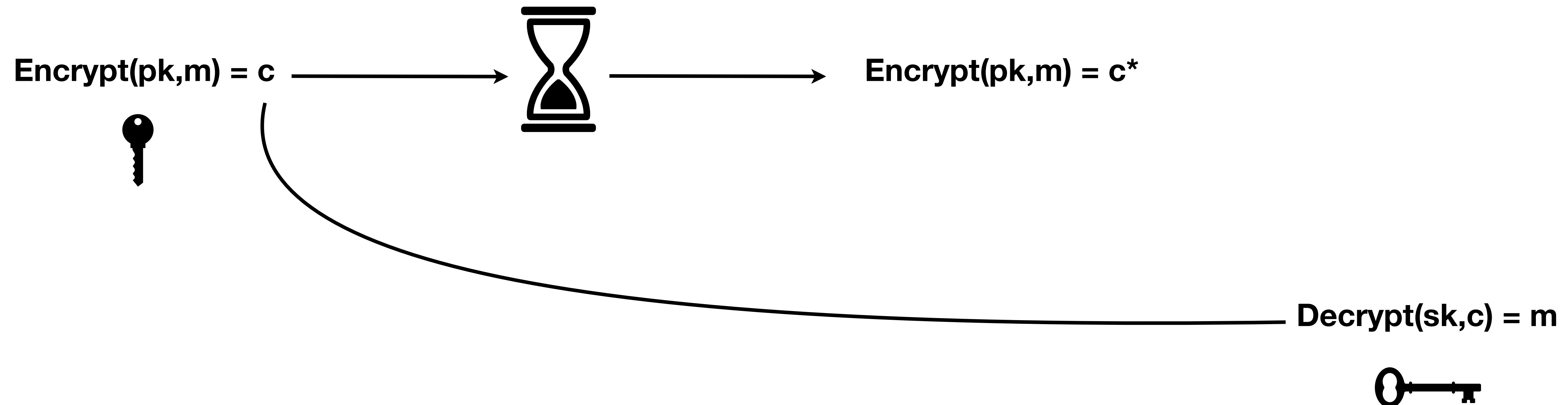
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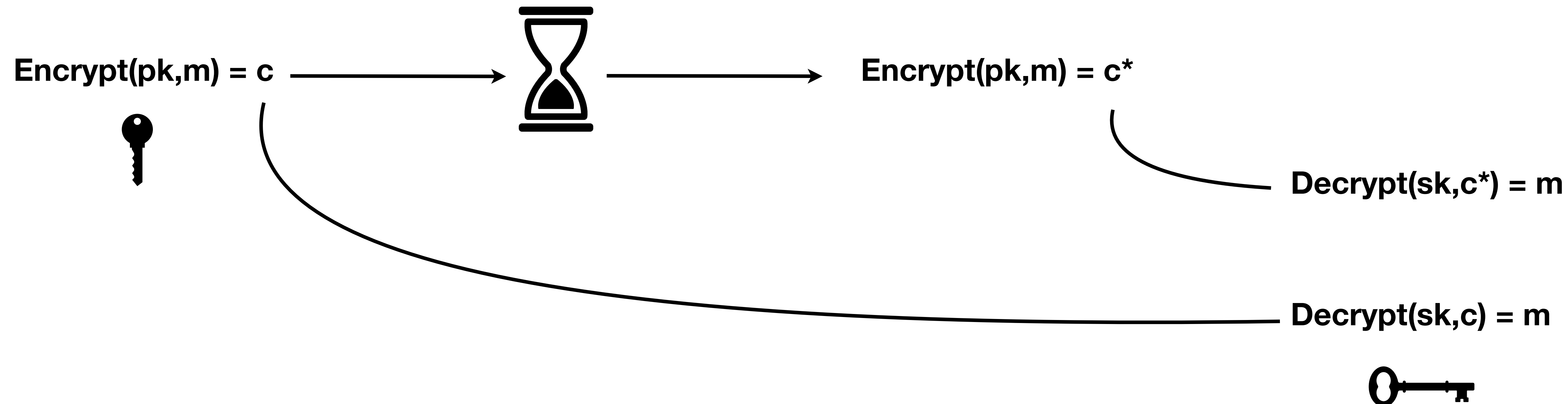
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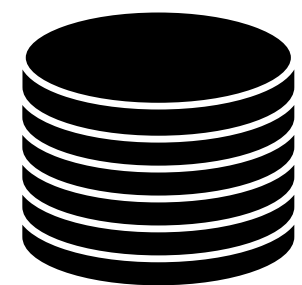
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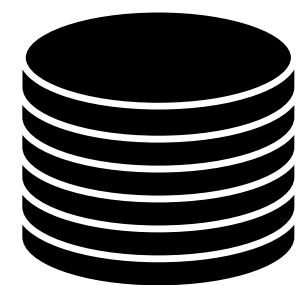


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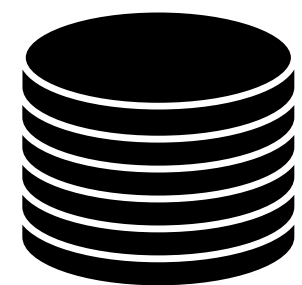


Key:  
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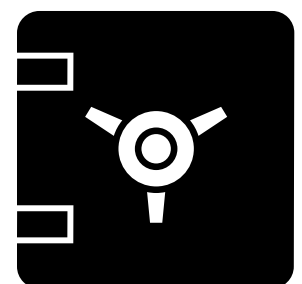
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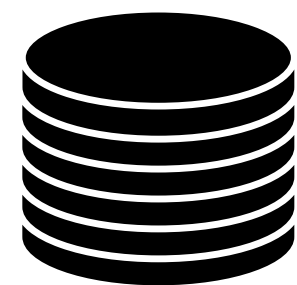


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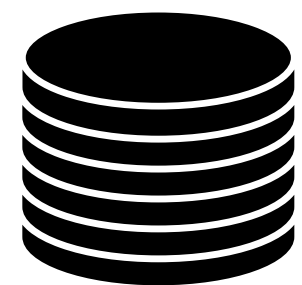


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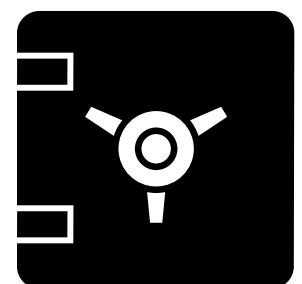
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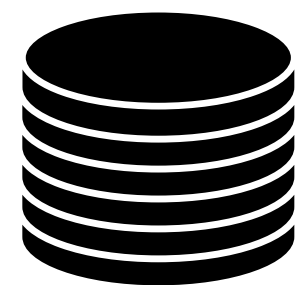


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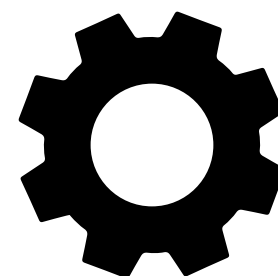
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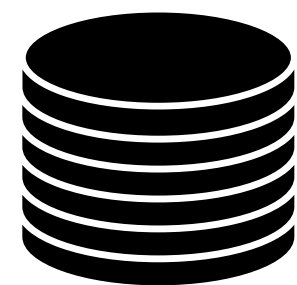
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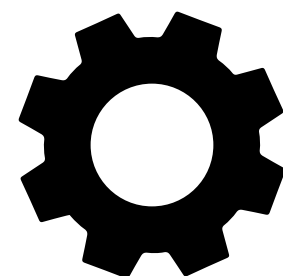
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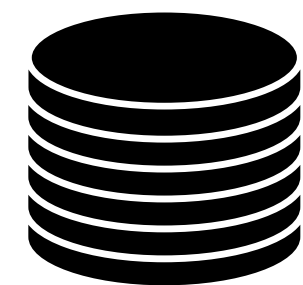
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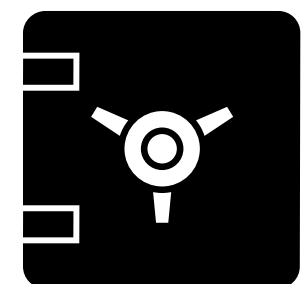
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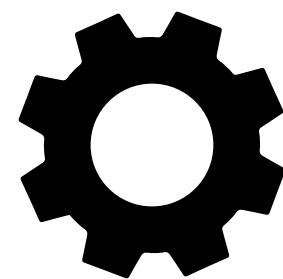
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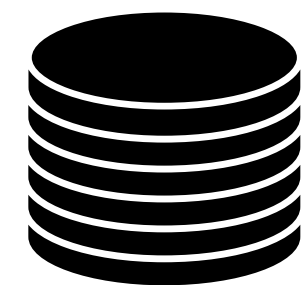
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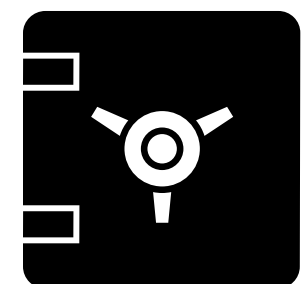
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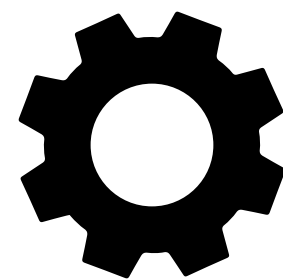
Key:  
 $p$ , odd integer



Cyphertext space:  
integers



Encryption function:  
 $c = pq + 2r + m$   
( $q, r$  random)



Decryption function:  
 $(c \bmod p) \bmod 2$

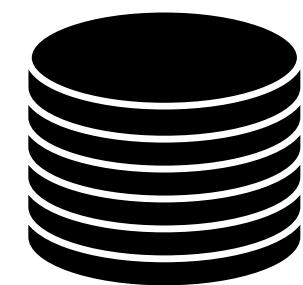
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A SHE scheme can homomorphically  
evaluate functions  $f$  with a limited depth

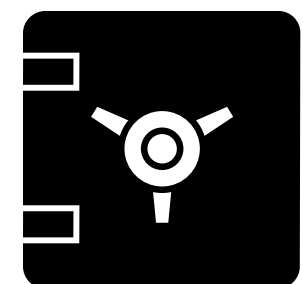
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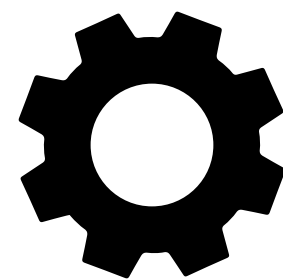
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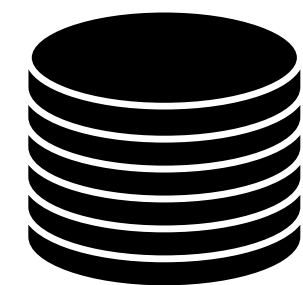
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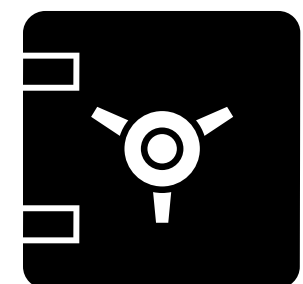
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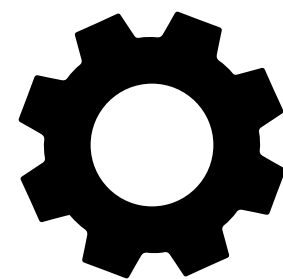
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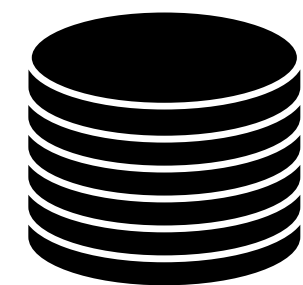
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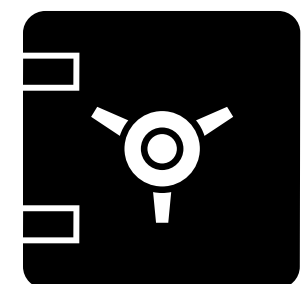
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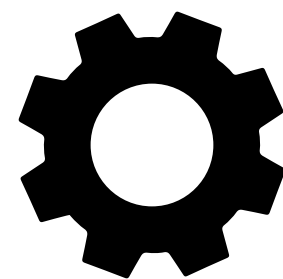
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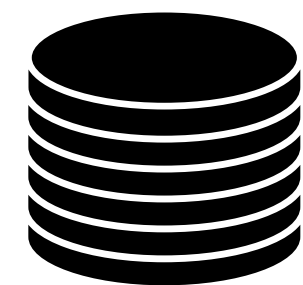
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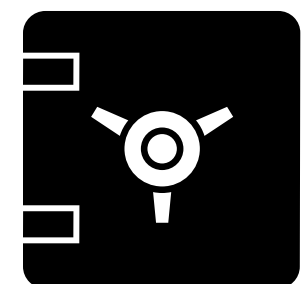
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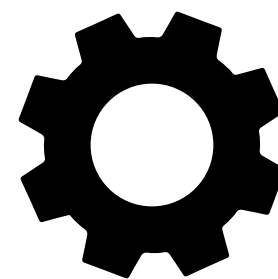
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If we add/multiply two cyphertext  $c1$  and  $c2$ ,  
the noise will be added/multiplied too

# Bootstrapping

Let's suppose we have a cyphertext  $c$   
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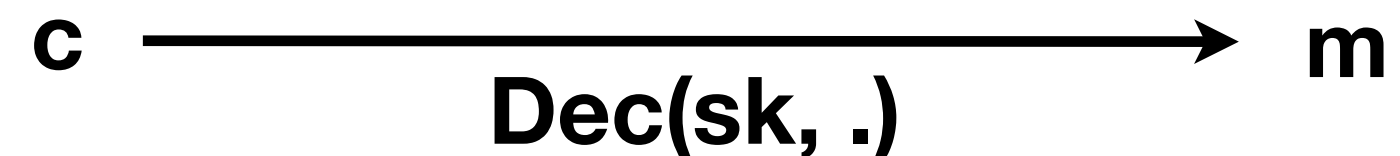




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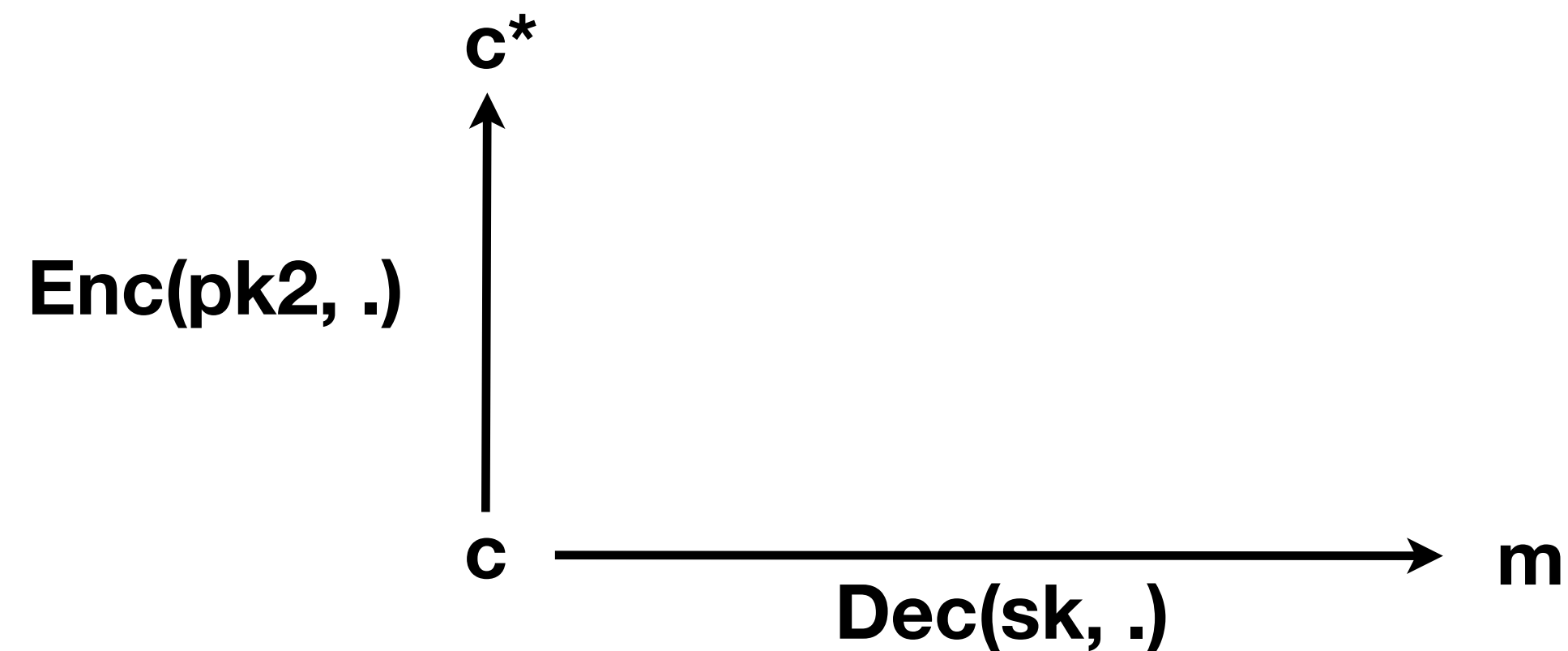
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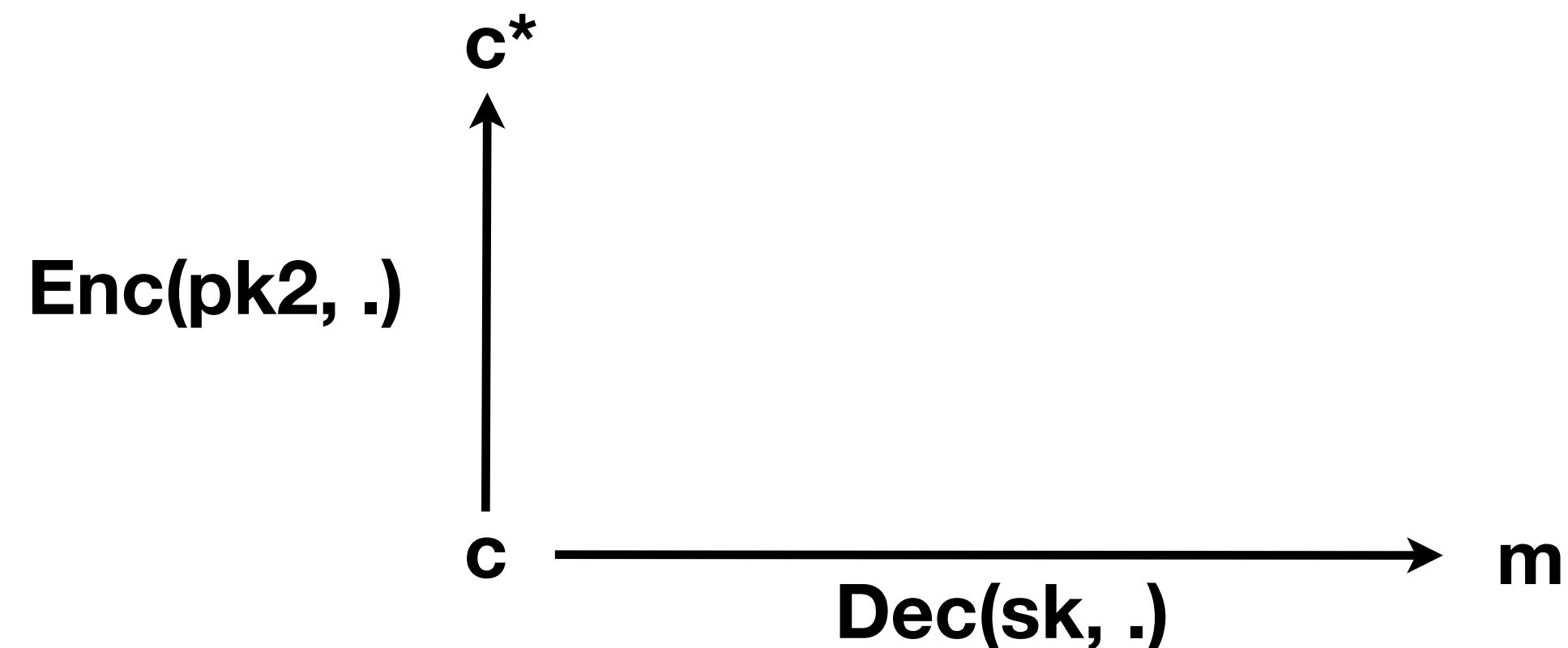


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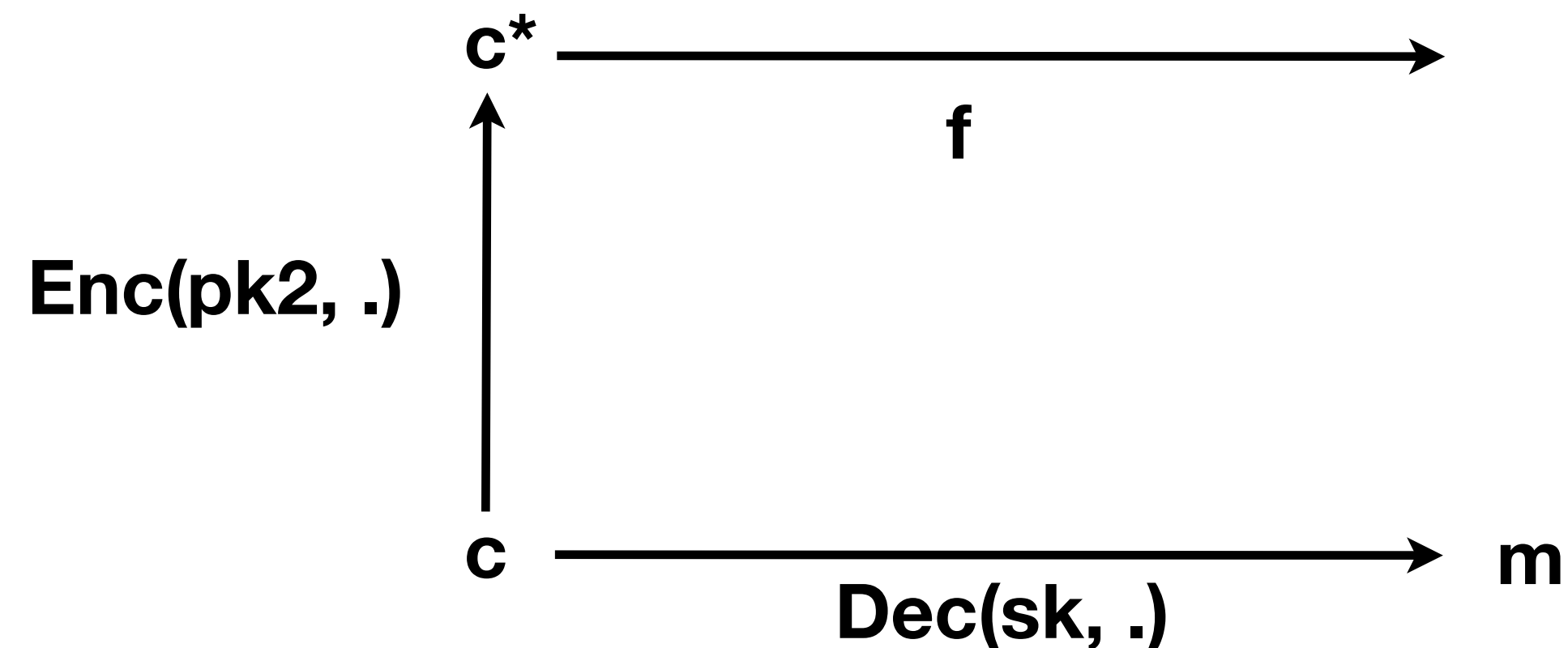
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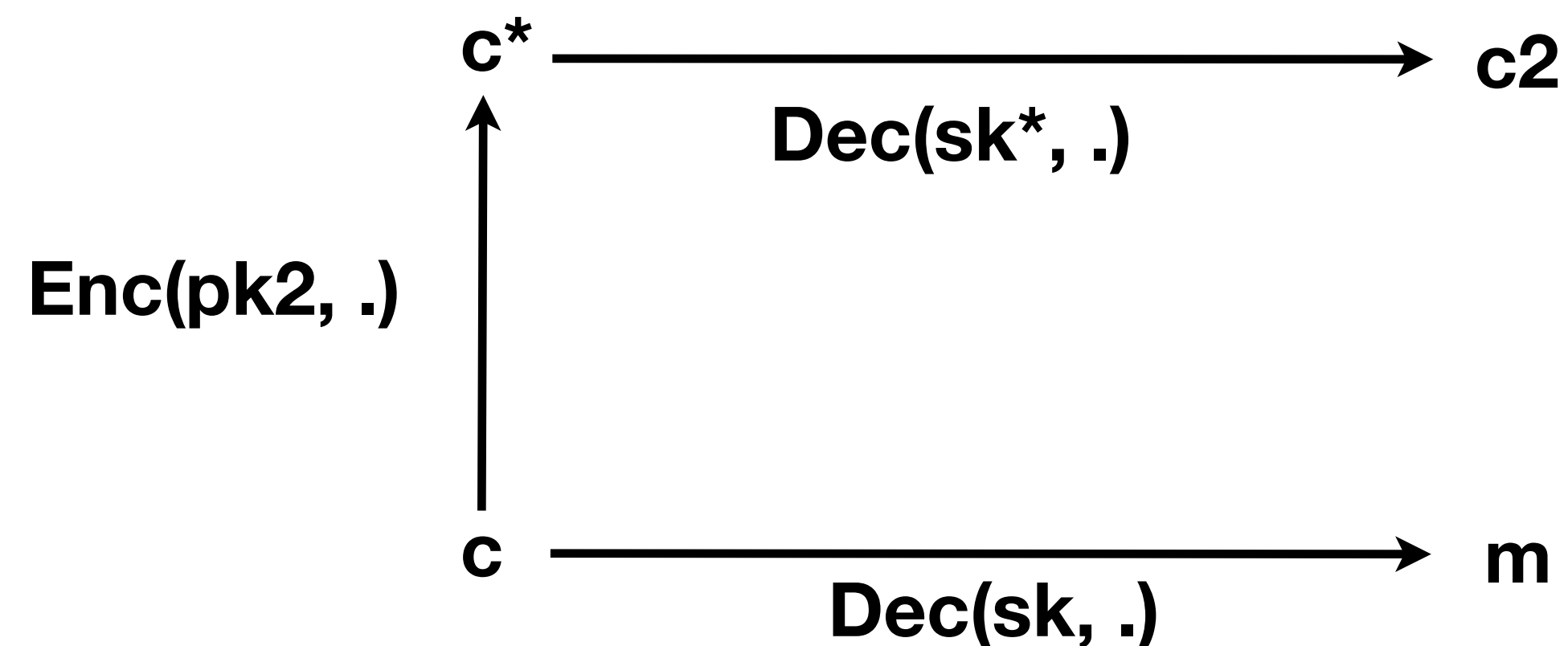
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$pk2$  = second public key  
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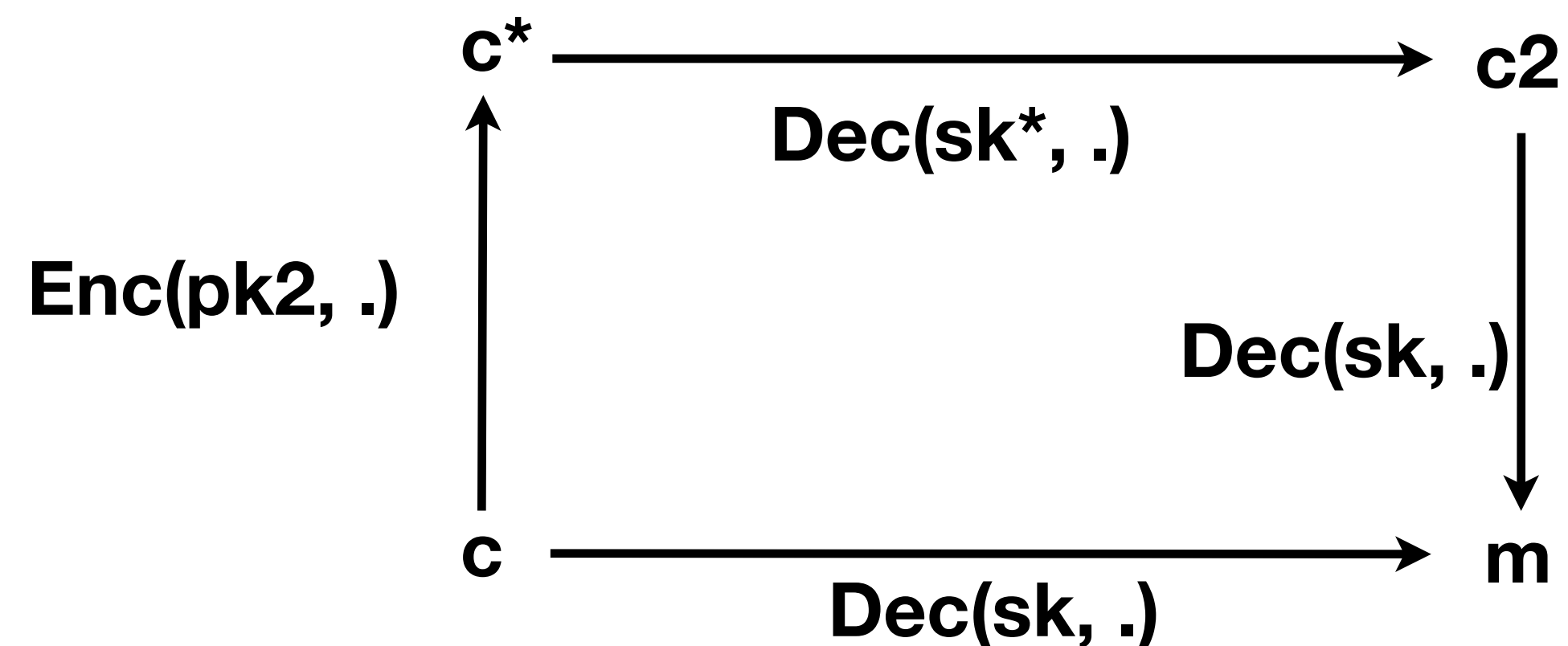
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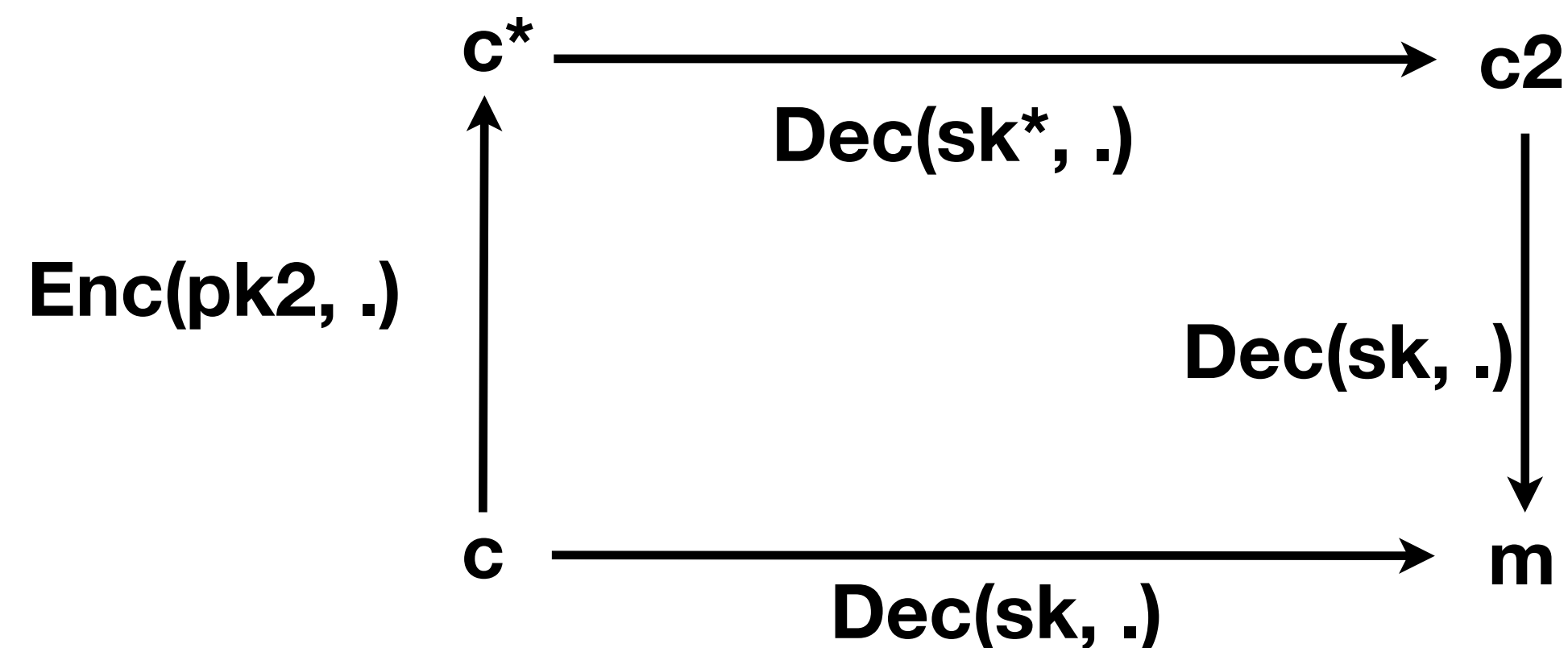
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A SHE scheme is called bootstrappable,  
if the Decryption function is a permitted function

# Literature overview

## 4 Generations of FHE

1st Gen.	C. Gentry, 2009	FHE Using Ideal Lattices	Solved 1978 Rivest Problem
1st Gen.	V. Dijk, C. Gentry, S. Halevi, V. Vaikuntanathan (DGHV, 2010)	FHE Over the Integers	Simplification of 2009 scheme
2nd Gen.	Z. Brakerski, Fan, Vercauteren (BFV, 2012)	Somewhat Practical FHE	More efficient SHE
3rd Gen.	Gentry, A. Sahai, and B. Waters (GSW, 2013)	HE from LWE: Conceptually-Simpler, Asymptotically-Faster	Faster bootstrapping, more efficient multiplications
4th Gen.	Cheon, Kim, Kim and Song (CKKS, 2016)	HE for arithmetic of approximate numbers	Built over complex/real numbers



# Calendar

MONTH	TASK
Mar - Jun	Mapped existing papers and studied Gentry's work (1st gen.)
July	Study next generation main papers/Map existing library's/packages
Ago - Set	Implementation of private regression models, tests and benchmarks
Out - Nov	Write

# References

**Rivest, 1978 - ON DATA BANKS AND PRIVACY HOMOMORPHISMS**

**Gentry, 2009 - Fully Homomorphic Encryption using Ideal Lattices**

**Gentry, 2010 - Fully Homomorphic Encryption over the Integers**