

# Multi-user Verifiable/Secure Computation

Ko Dokmai, Saba Eskandarian

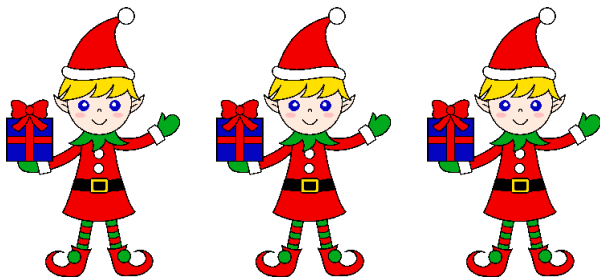
# Goals

Extend cryptographic protocols with applications to cloud computing from a two-party to a multi-party setting:

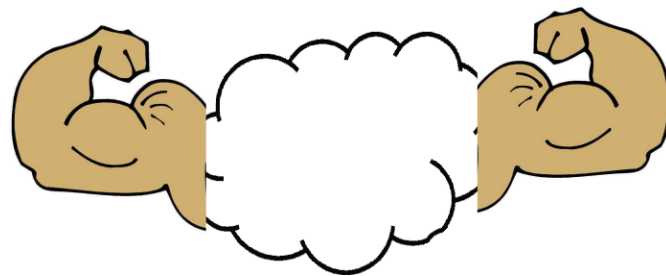
- Verifiable Computation
- Secure Computation

# Common Setting

Weak Clients



Powerful Cloud

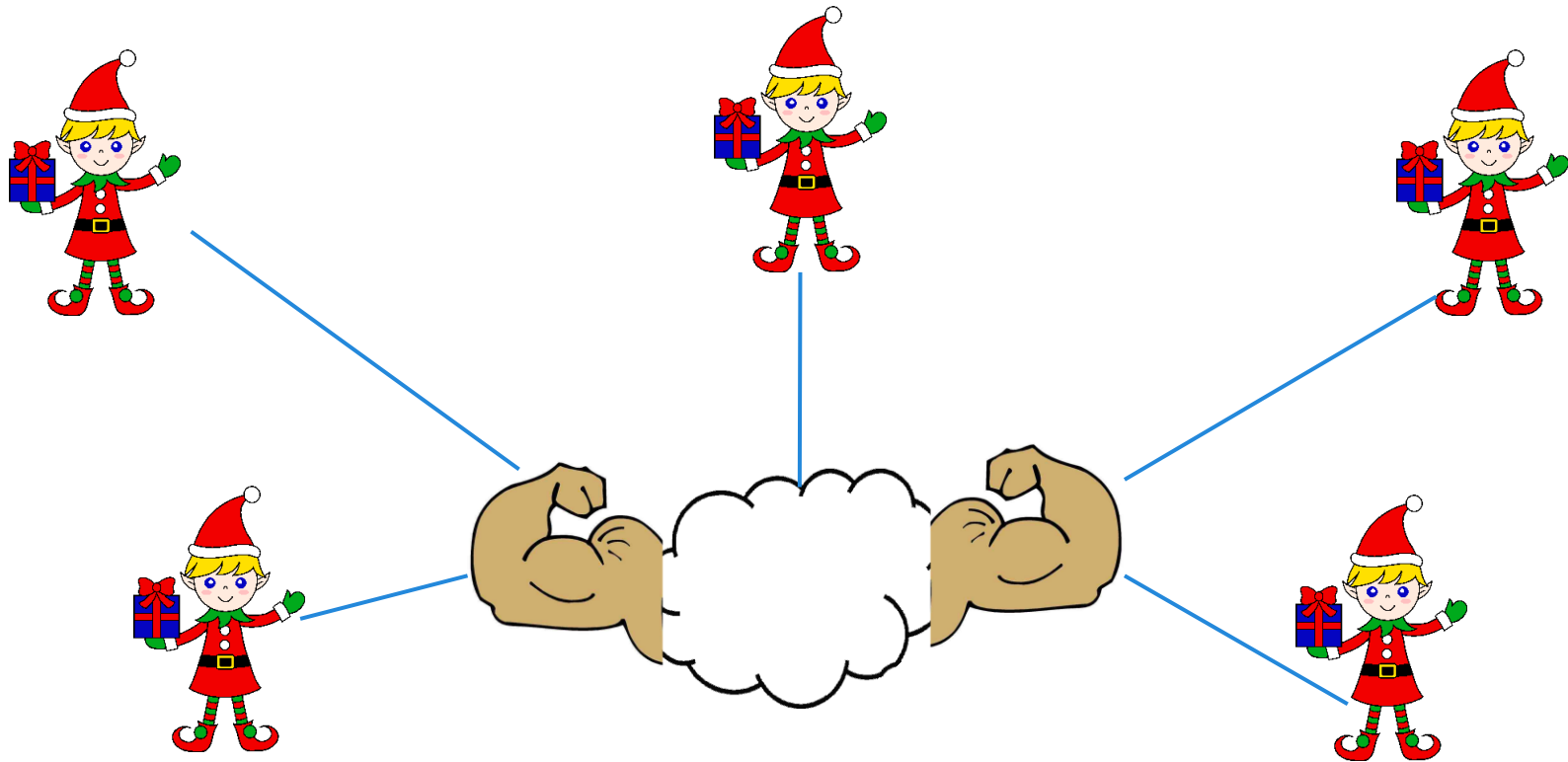


# Common Elements

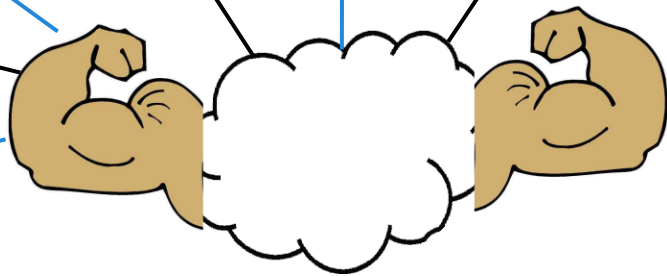
- Multiclient setting
- Outsourced computation
- Best definitions = VBB Obfuscation

# Multi-user Verifiable Computation

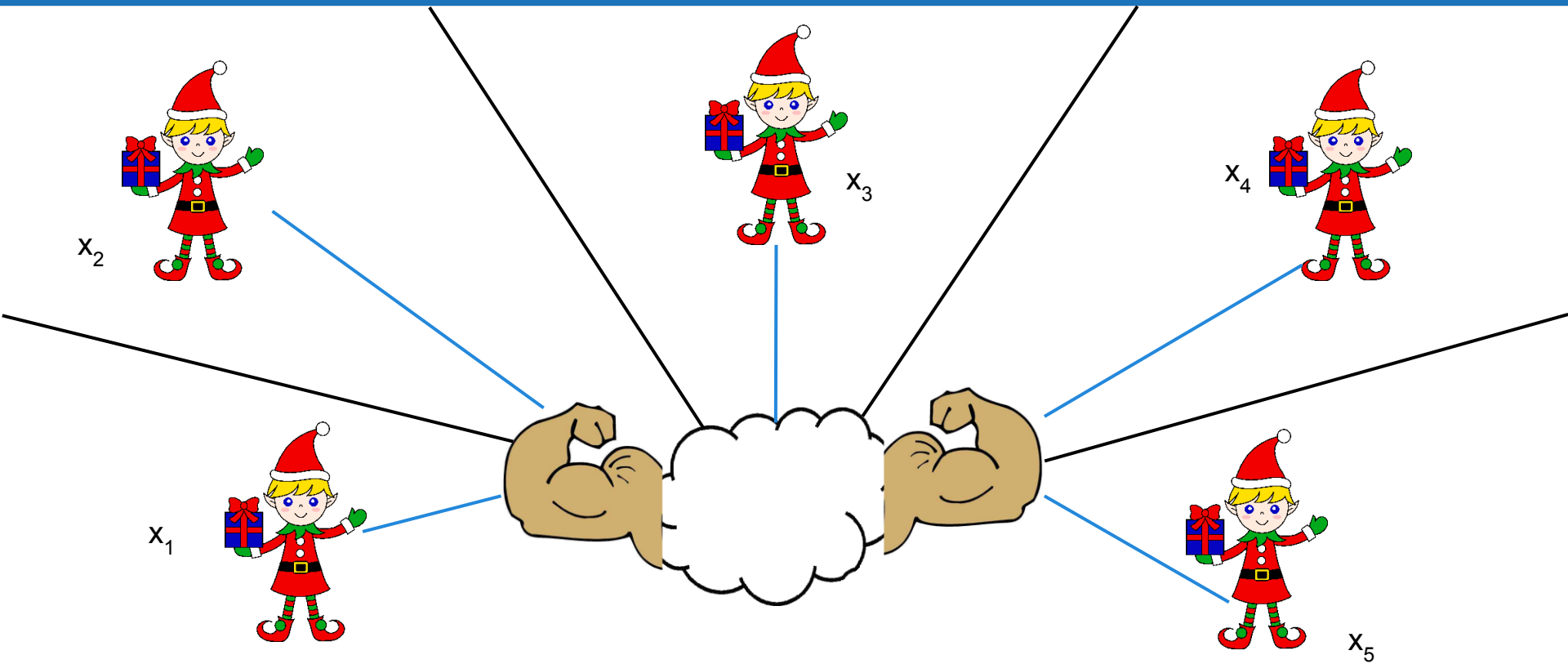
# Setting



# Setting

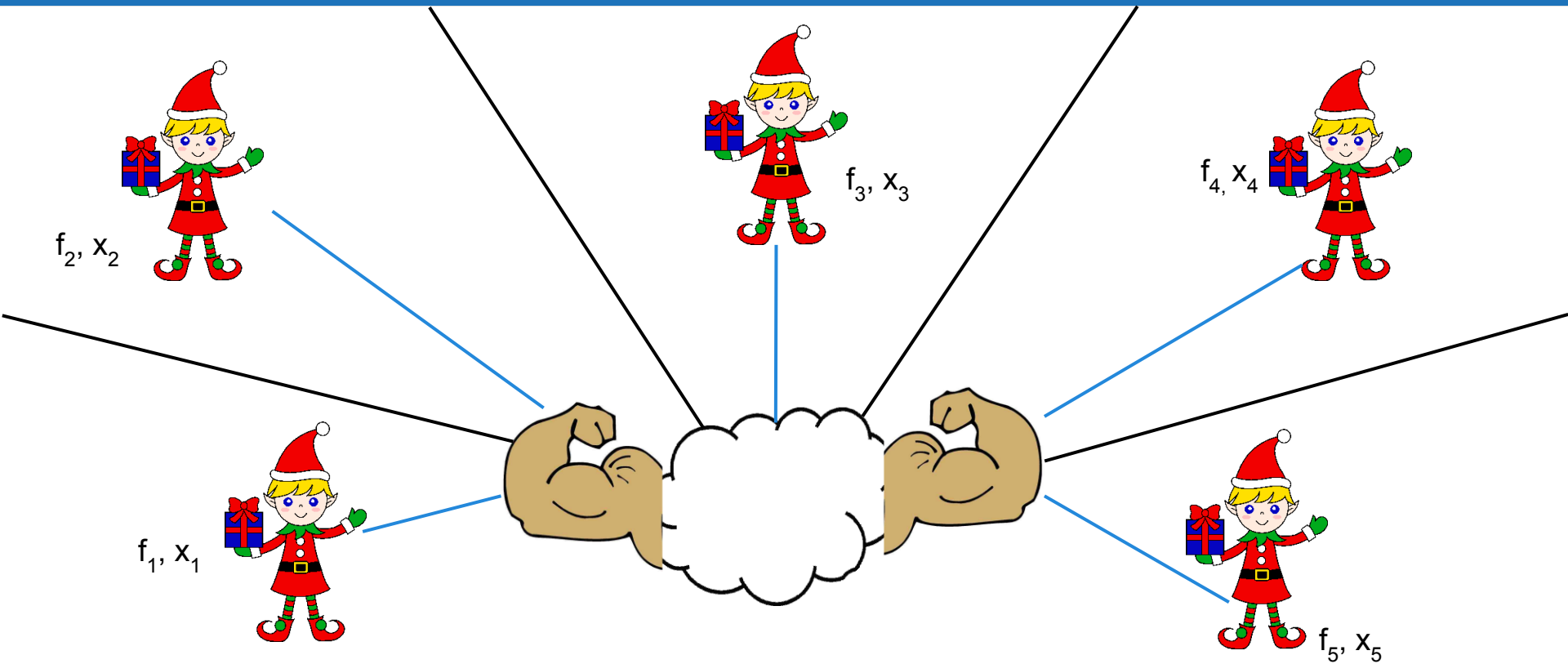


# Setting

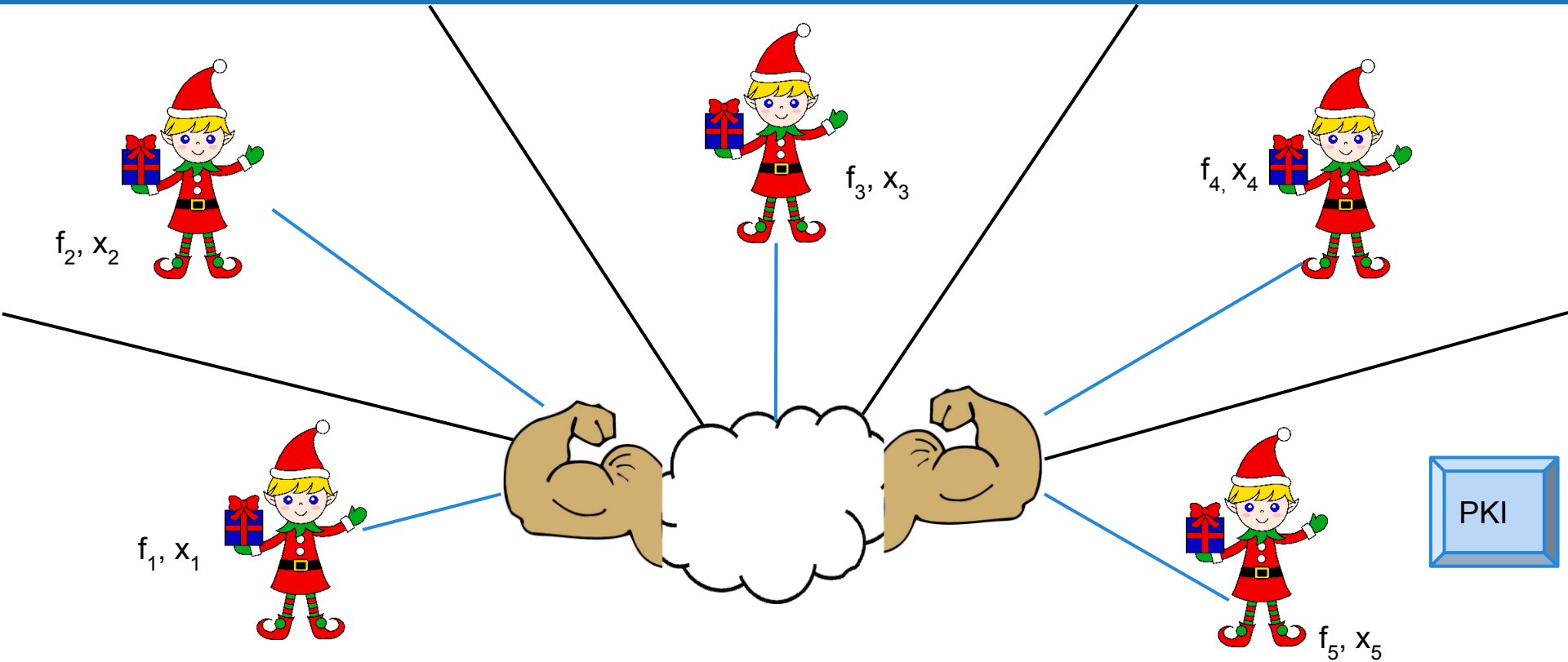




# Setting



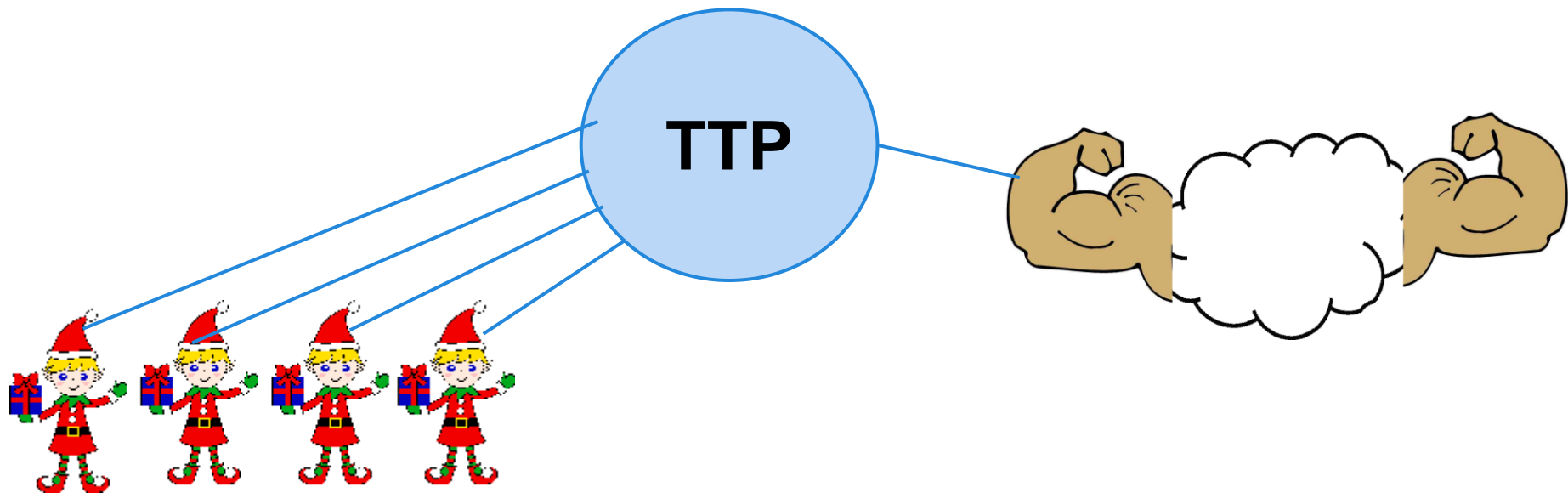
# Setting



# Security

- Real/Ideal paradigm
- Most desirable definition is unattainable (implies VBB)
- Ideal definition behavior differs based on who is corrupted
- we will assume semi-honest cloud and clients

# Ideal Situation



# Primitives Used

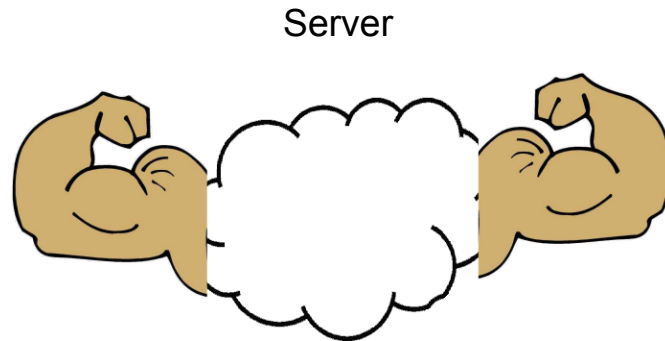
- Proxy oblivious transfer (POT)
- Garbled Circuits
- Fully homomorphic encryption (FHE)

# Garbled Circuits

- $\text{Garble}(1^k, C) \rightarrow \text{pk}, \text{sk}, \Gamma$
- $\text{Genc}(\text{pk}, x) \rightarrow c$
- $\text{Geval}(\Gamma, c) \rightarrow Y$
- $\text{Gdec}(\text{sk}, Y) \rightarrow y$
  
- $y = C(x)$
- Extra requirements

# 1st try: single-client, single-use

Intuition:  $P_1$  Garbles function and input,  
Server evaluates and sends back results.



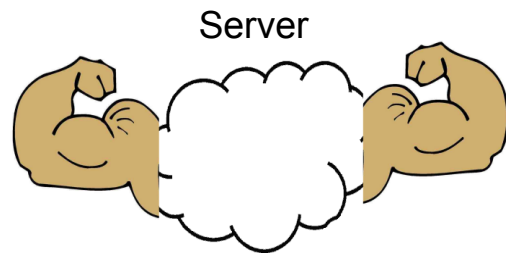
# 1st try: single-client, single-use



$\text{Garble}(1^k, f) \rightarrow \text{pk}, \text{sk}, \Gamma$

$\text{Genc}(\text{pk}, x) \rightarrow c$

$\Gamma, c$





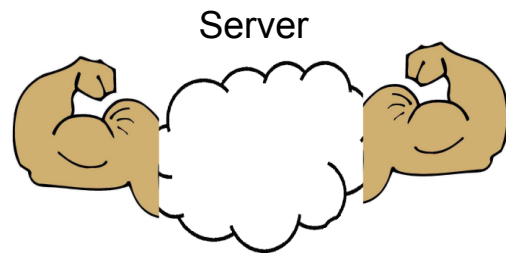
# 1st try: single-client, single-use



$\text{Garble}(1^k, f) \rightarrow \text{pk}, \text{sk}, \Gamma$

$\text{Genc}(\text{pk}, x) \rightarrow c$

$\Gamma, c$



$\text{Geval}(\Gamma, c) \rightarrow Y$

$Y$

# 1st try: single-client, single-use



$\text{Garble}(1^k, f) \rightarrow \text{pk}, \text{sk}, \Gamma$

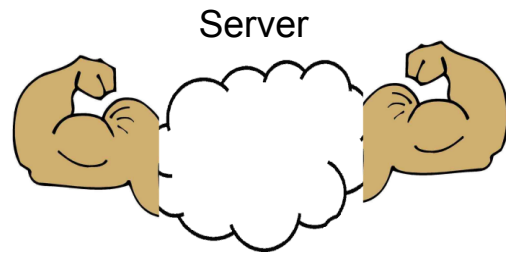
$\text{Genc}(\text{pk}, x) \rightarrow c$

$\Gamma, c$

$\text{Geval}(\Gamma, c) \rightarrow Y$

$Y$

$\text{Gdec}(\text{sk}, Y) \rightarrow y$



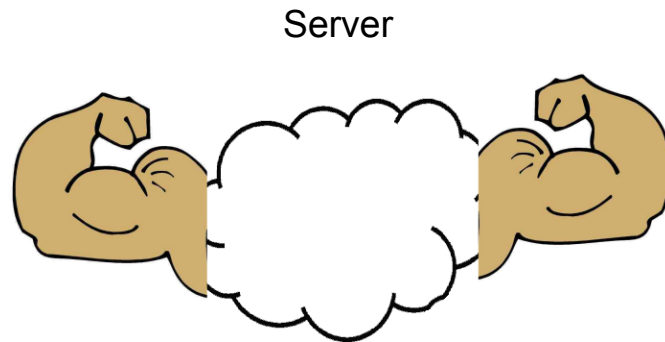
Problem: 1<sup>st</sup> try only works once.

How can we allow for multiple  
function evaluations?

Why not just keep sending more inputs?

## 2nd try: single-client, multi-use

Intuition: garbled inputs encrypted with FHE

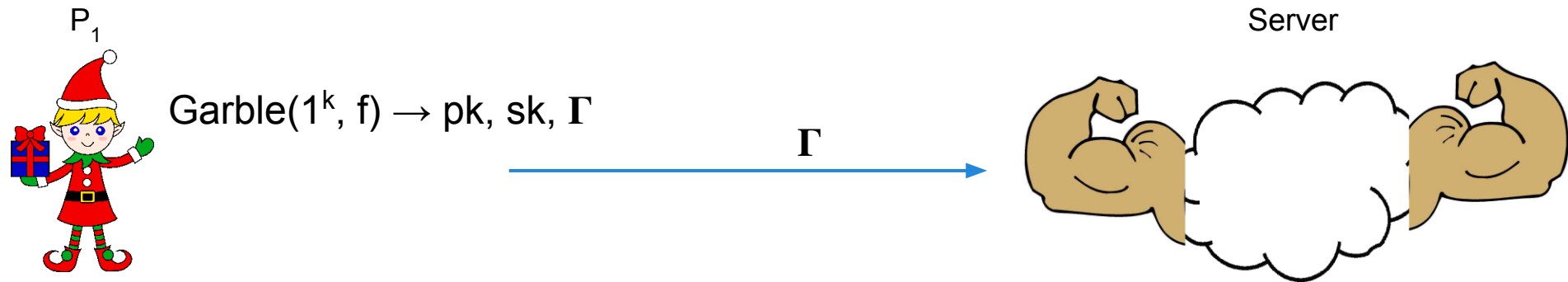


# Fully Homomorphic Encryption

- $\text{Fgen}(1^k) \rightarrow (\text{pk}_{\text{FHE}}, \text{sk}_{\text{FHE}})$
- $\text{Fenc}(\text{pk}_{\text{FHE}}, m) \rightarrow c$
- $\text{Fdec}(\text{sk}_{\text{FHE}}, c') \rightarrow m'$
- $\text{Feval}(\text{pk}_{\text{FHE}}, f, c_1, \dots, c_n)$

# 2nd try: single-client, multi-use

## Setup Phase



# 2nd try: single-client, multi-use

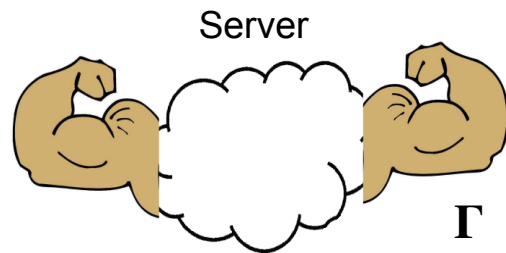


$\text{Fgen}(1^k) \rightarrow (\text{pk}_{\text{FHE}}, \text{sk}_{\text{FHE}})$

$\text{Genc}(\text{pk}, x) \rightarrow c$

$\text{Fenc}(\text{pk}_{\text{FHE}}, c) \rightarrow c'$

$c'$



# 2nd try: single-client, multi-use



$P_1$

$\text{Fgen}(1^k) \rightarrow (\text{pk}_{\text{FHE}}, \text{sk}_{\text{FHE}})$

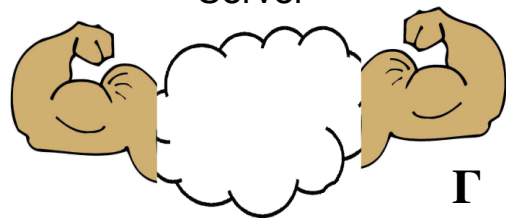
$\text{Genc}(\text{pk}, x) \rightarrow c$

$\text{Fenc}(\text{pk}_{\text{FHE}}, c) \rightarrow c'$

$c'$

$Y'$

Server



$\Gamma$

$\text{Feval}(\text{pk}_{\text{FHE}}, \text{Geval}(\Gamma, \cdot), c') \rightarrow Y'$



# 2nd try: single-client, multi-use



P<sub>1</sub>

$\text{Fgen}(1^k) \rightarrow (\text{pk}_{\text{FHE}}, \text{sk}_{\text{FHE}})$

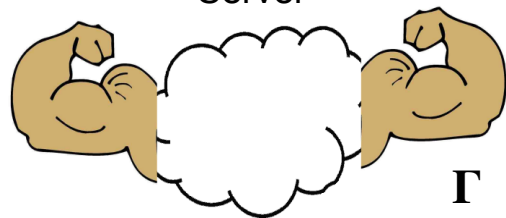
$\text{Genc}(\text{pk}, x) \rightarrow c$

$\text{Fenc}(\text{pk}_{\text{FHE}}, c) \rightarrow c'$

$c'$

$Y'$

Server



$\Gamma$

$\text{Feval}(\text{pk}_{\text{FHE}}, \text{Geval}(\Gamma, \cdot), c') \rightarrow Y'$

# 2nd try: single-client, multi-use



$P_1$

$\text{Fgen}(1^k) \rightarrow (\text{pk}_{\text{FHE}}, \text{sk}_{\text{FHE}})$

$\text{Genc}(\text{pk}, x) \rightarrow c$

$\text{Fenc}(\text{pk}_{\text{FHE}}, c) \rightarrow c'$

$c'$

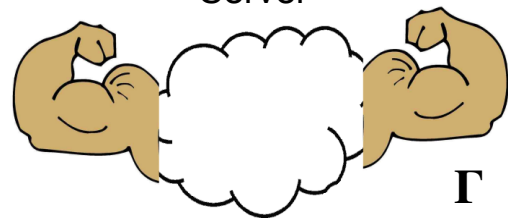
$\text{Feval}(\text{pk}_{\text{FHE}}, \text{Geval}(\Gamma, \cdot), c') \rightarrow Y'$

$Y'$

$\text{Fdec}(\text{sk}_{\text{FHE}}, Y') \rightarrow Y$

$\text{Gdec}(\text{sk}, Y) \rightarrow y$

Server



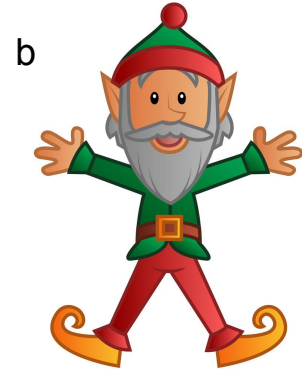
$\Gamma$

How to go multiparty?

We'll need some new tools.

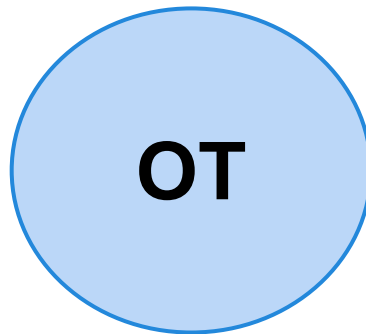
What prevents the 2<sup>nd</sup> try from generalizing to multiparty setting?

# Oblivious Transfer (OT)

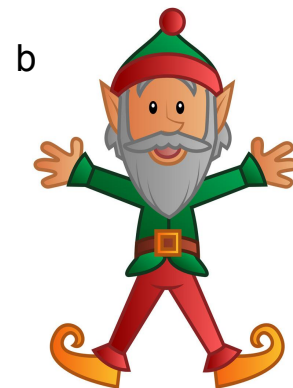


# Oblivious Transfer (OT)

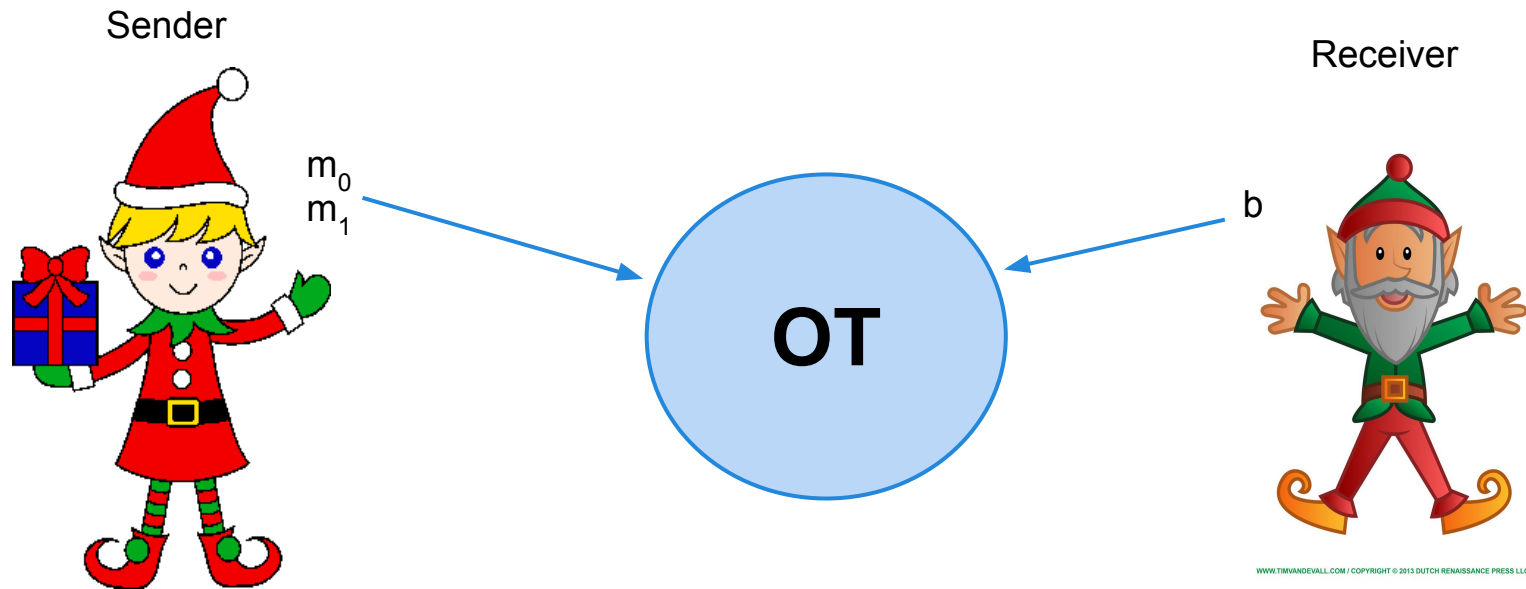
Sender



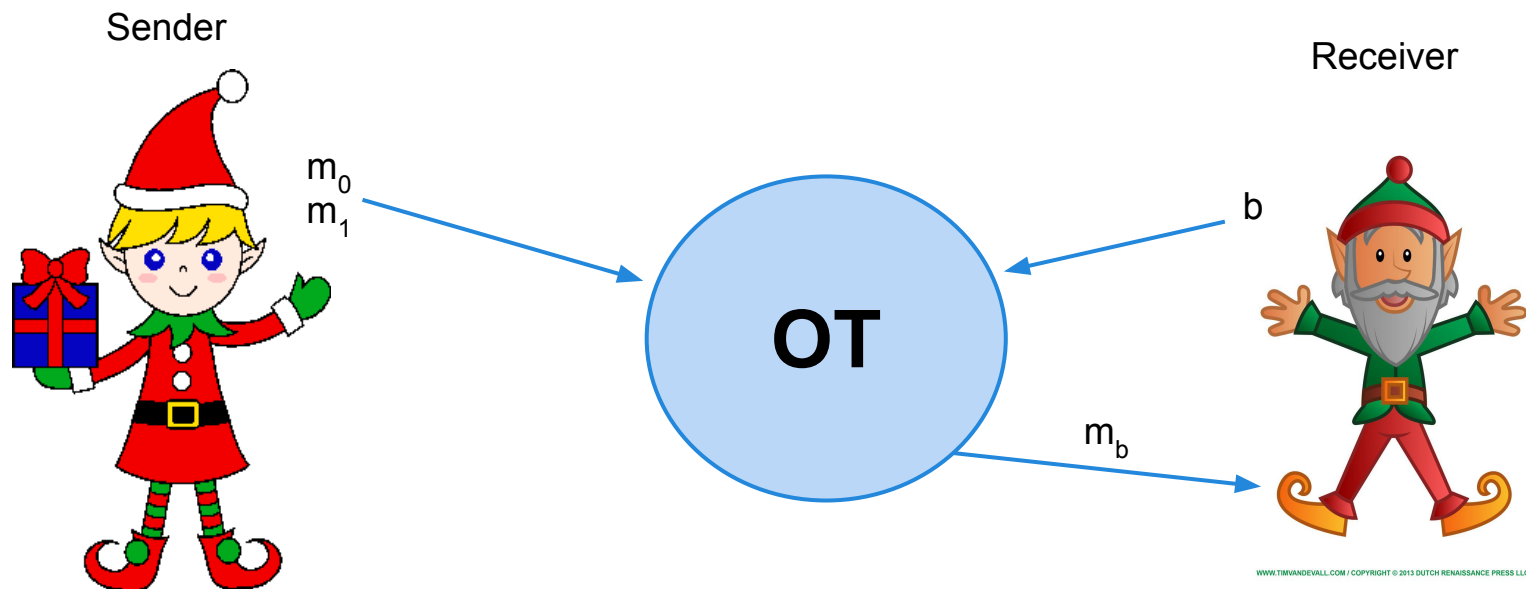
Receiver



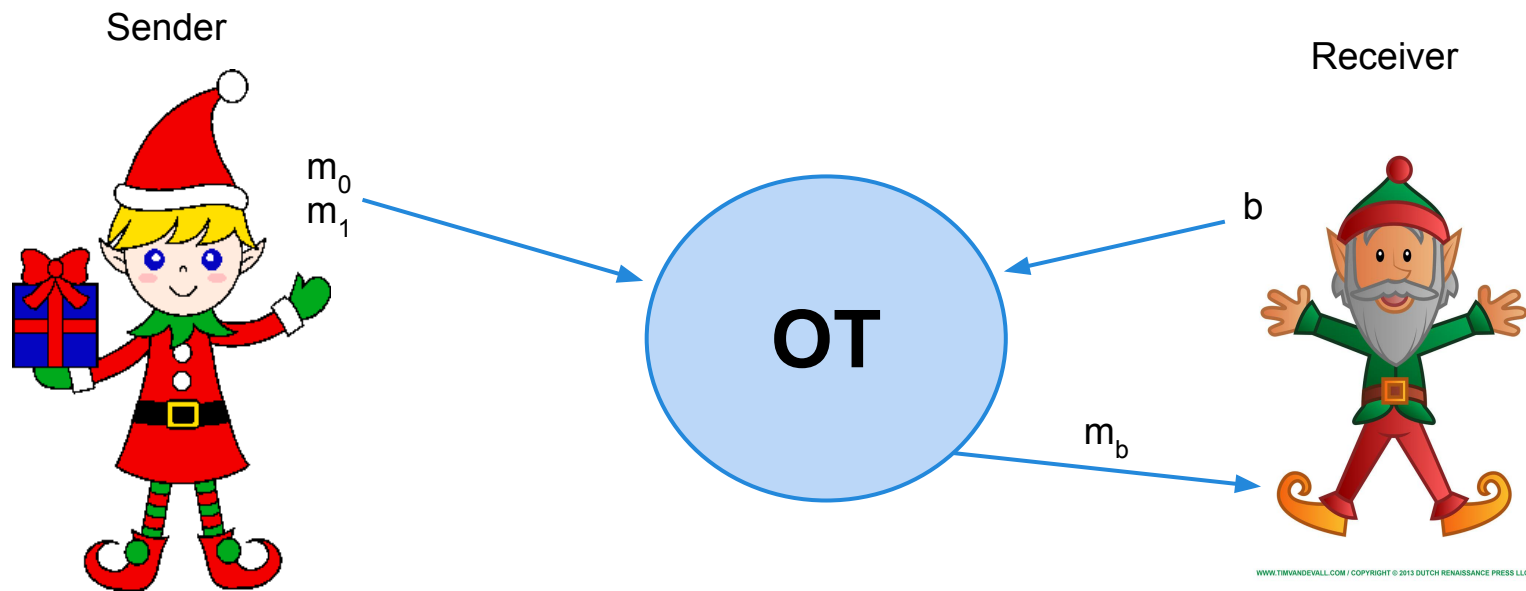
# Oblivious Transfer (OT)



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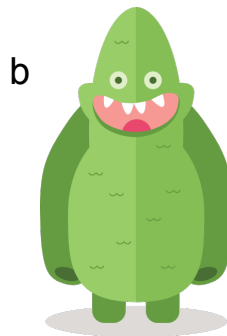


- does not learn  $b$

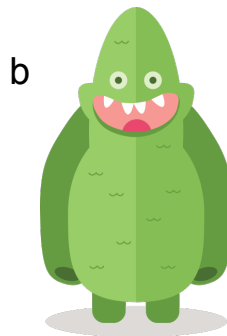
- learns  $m_b$
- does not learn  $m_{1-b}$



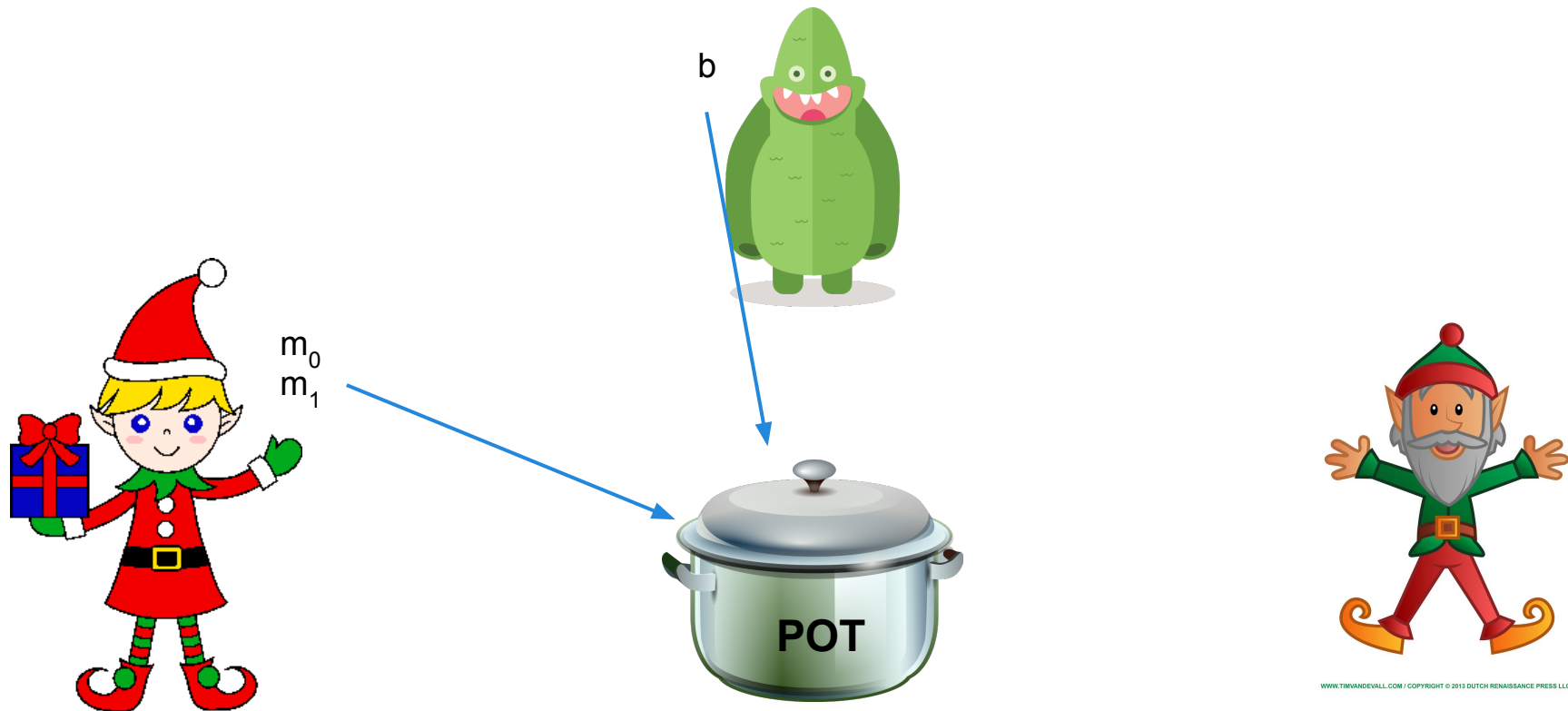
# Proxy Oblivious Transfer (POT)



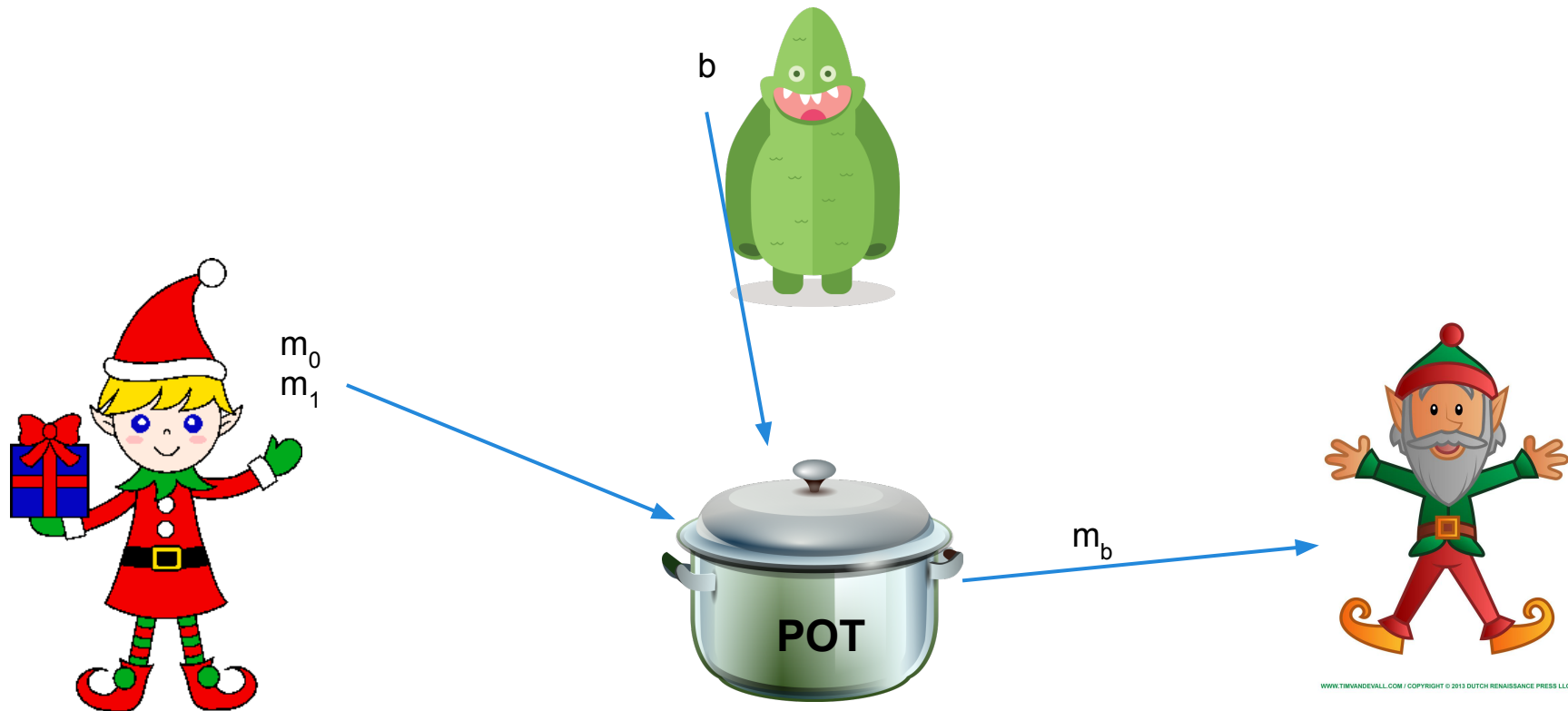
# Proxy Oblivious Transfer (POT)



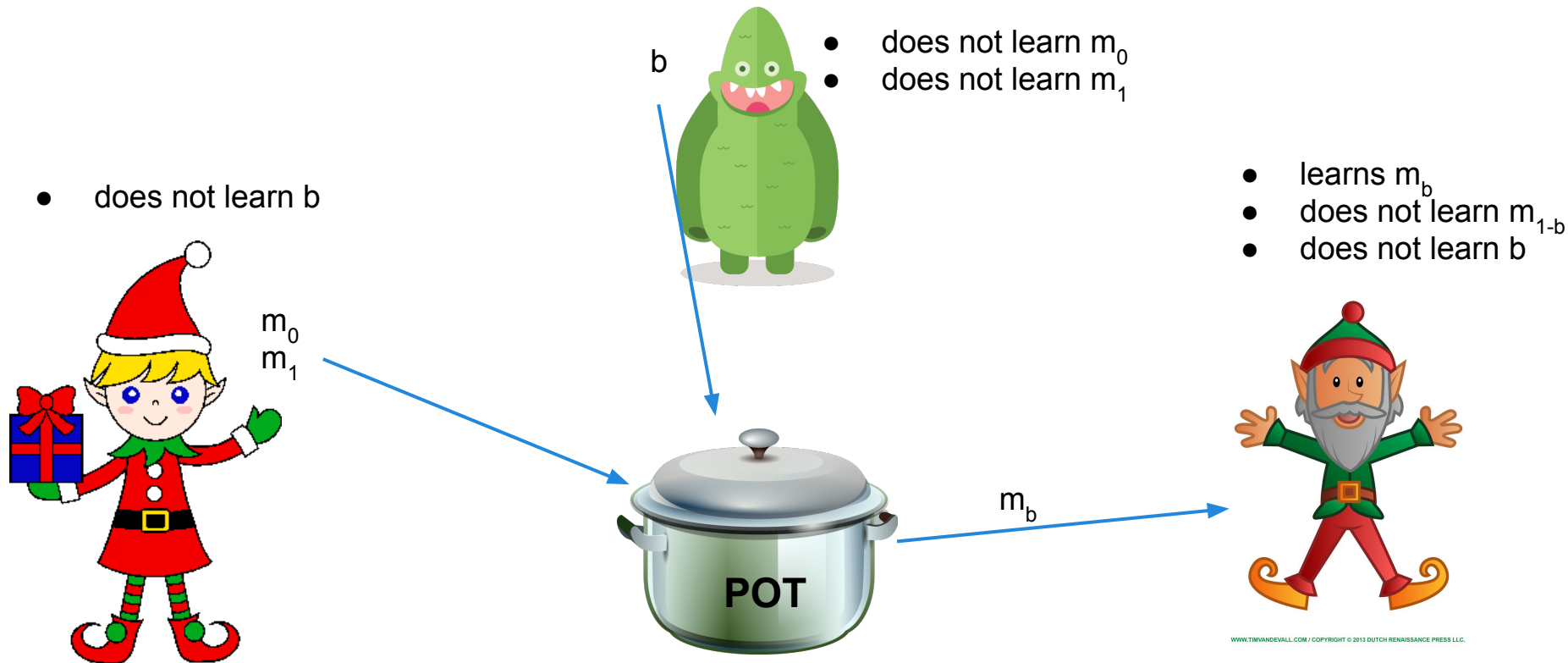
# Proxy Oblivious Transfer (POT)



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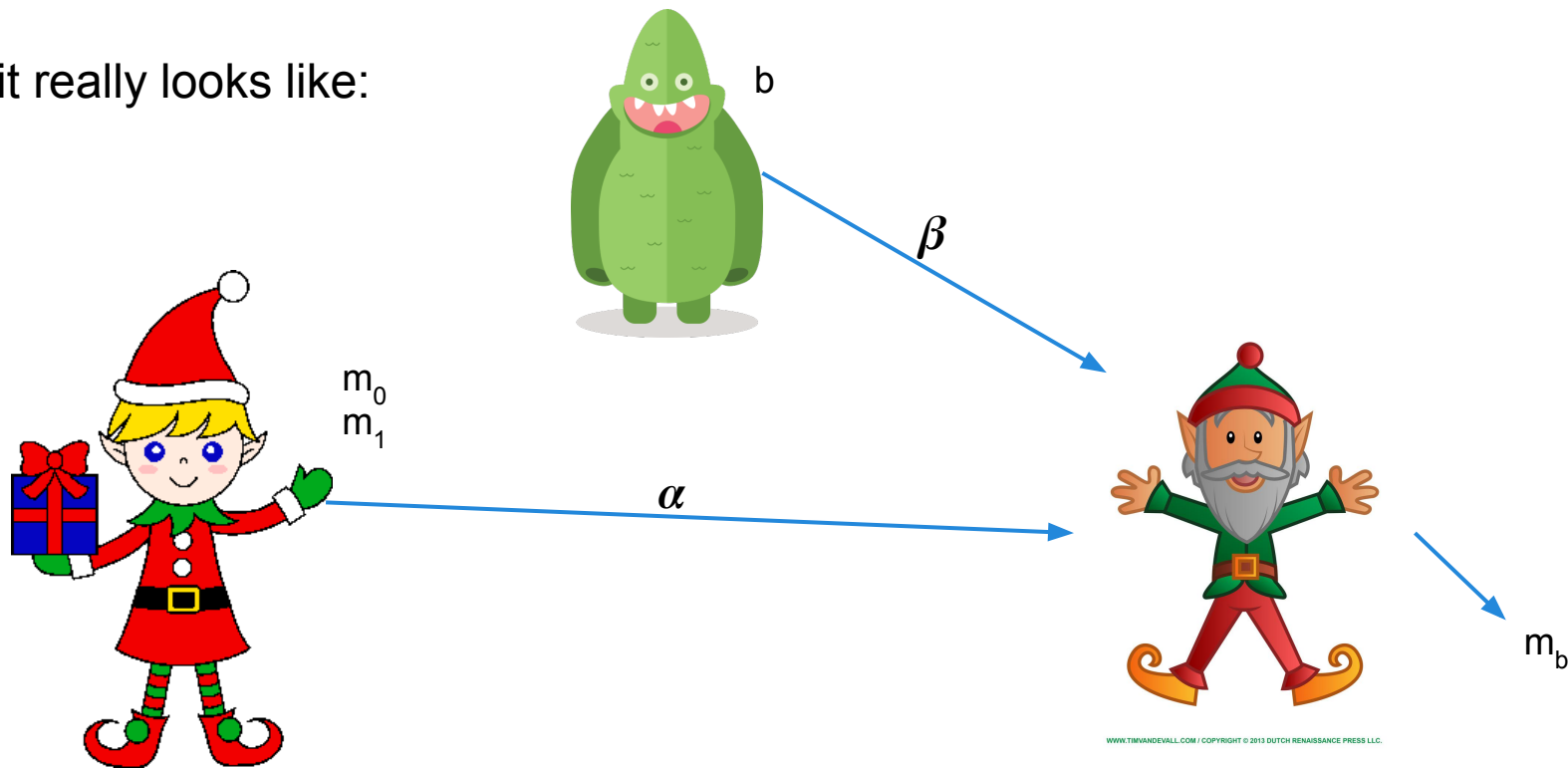


# Proxy Oblivious Transfer (POT)



# Proxy Oblivious Transfer (POT)

What it really looks like:

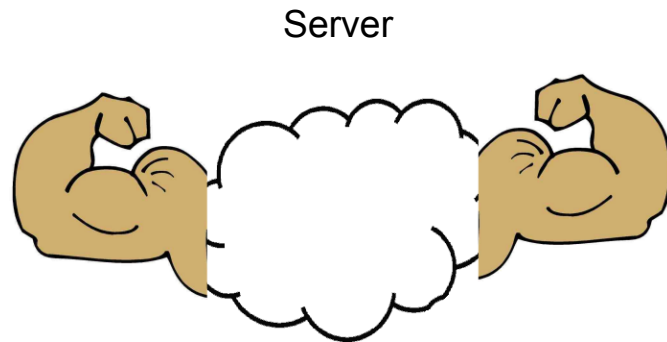
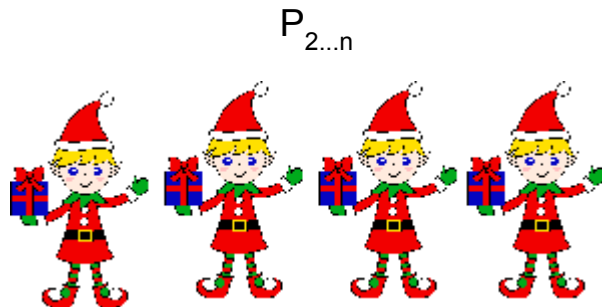


# Proxy Oblivious Transfer

- $\text{SetupS}(1^k) \rightarrow (\text{pk}_s, \text{sk}_s)$
- $\text{SetupC}(1^k) \rightarrow (\text{pk}_c, \text{sk}_c)$
- $\text{Snd}(\text{pk}_c, \text{sk}_s, m_0, m_1) \rightarrow \alpha$
- $\text{Chs}(\text{pk}_s, \text{sk}_c, b) \rightarrow \beta$
- $\text{Prx}(\text{pk}_s, \text{pk}_c, \alpha, \beta) \rightarrow y$
- $y = m_b$

# 3rd try: multi-user, multi-use

Intuition: use POT to allow  $P_2 \dots P_n$  to pick their inputs for the Garbled Circuit





# 3rd try: multi-user, multi-use

## Setup Phase

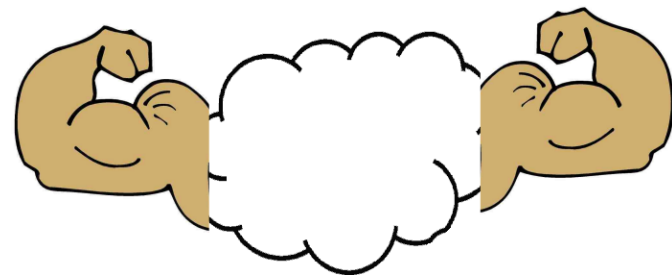


$P_1$

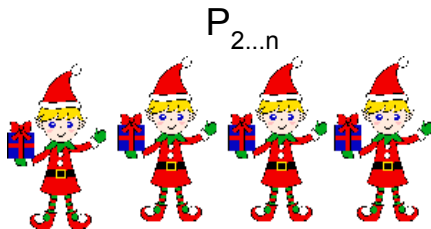
$\text{Garble}(1^k, f) \rightarrow \text{pk}, \text{sk}, \Gamma$

$\text{SetupS}(1^k) \rightarrow \text{pk}_s, \text{sk}_s$

$\Gamma$



Server



$P_{2...n}$

$\text{SetupC}(1^k) \rightarrow \text{pk}_c, \text{sk}_c$

# 3rd try: multi-user, multi-use



$$\text{Fgen}(1^k) \rightarrow (\text{pk}_{\text{FHE}}, \text{sk}_{\text{FHE}})$$

$$\text{Genc}(\text{pk}, x_1) \rightarrow c_1$$

$$\text{Fenc}(\text{pk}_{\text{FHE}}, c_1) \rightarrow c_1'$$

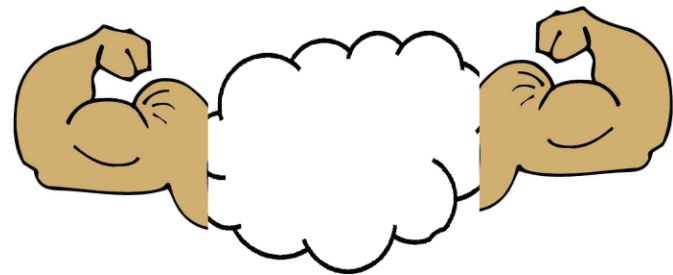
for  $i$  from 2 to  $n$  and for  $j$  from 1 to  $l$   
( $l$  = length of input):

$$x_0 = \text{Fenc}(\text{sk}_{\text{FHE}}, \text{Genc}(\text{pk}, 0))$$

$$x_1 = \text{Fenc}(\text{sk}_{\text{FHE}}, \text{Genc}(\text{pk}, 1))$$

$$\text{Snd}(\text{pk}_{\text{ci}}, \text{sk}_s, x_0, x_1) \rightarrow \alpha_{ij}$$

$c_1', \alpha$



# 3rd try: multi-user, multi-use

$P_{2...n}$

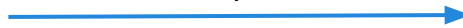


for  $j$  from 1 to  $l$ :

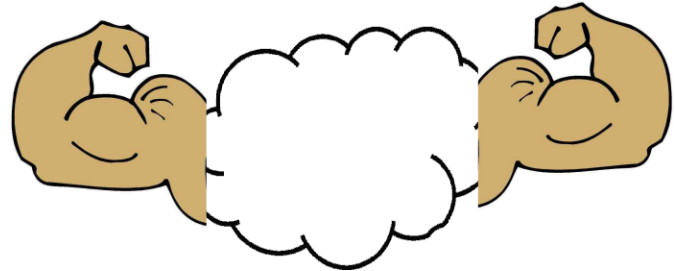
$\text{Chs}(\text{pk}_s, \text{sk}_{ci}, x_{ij}) \rightarrow \beta_{ij}$

Send all  $\beta_{ij}$  to Server

$\beta_i$



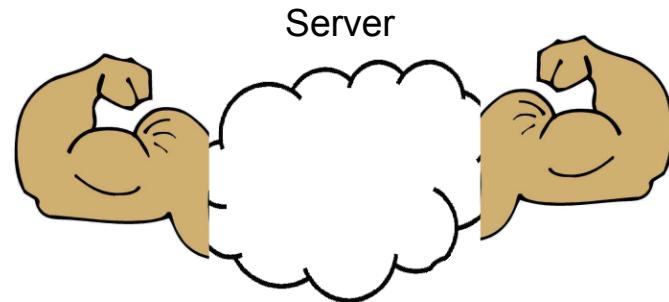
Server



# 3rd try: multi-user, multi-use



$Y'$



For each  $i,j$ :  $\text{Prx}(\text{pk}_s, \text{pk}_{c_i}, \alpha_{ij}, \beta_{ij}) \rightarrow c'_{ij}$   
Reassemble  $c'_1$  and all  $c'_{ij}$  in order of  
subscript into  $c'$

$\text{Feval}(\text{pk}_{\text{FHE}}, \text{Geval}(\Gamma, \cdot), c') \rightarrow Y'$

# 3rd try: multi-user, multi-use

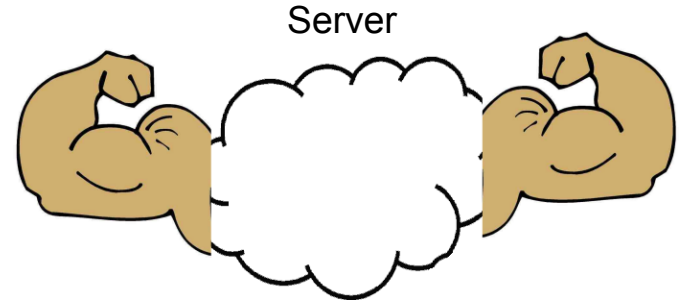


$P_1$

$Y'$

$$\text{Fdec}(\text{sk}_{\text{FHE}}, Y') \rightarrow Y$$

$$\text{Gdec}(\text{sk}, Y) \rightarrow y$$



Server

For each  $i, j$ :  $\text{Prx}(\text{pk}_s, \text{pk}_{c_i}, \alpha_{ij}, \beta_{ij}) \rightarrow c'_{ij}$   
Reassemble  $c'_1$  and all  $c'_{ij}$  in order of subscript into  $c'$

$$\text{Feval}(\text{pk}_{\text{FHE}}, \text{Geval}(\Gamma, \cdot), c') \rightarrow Y'$$

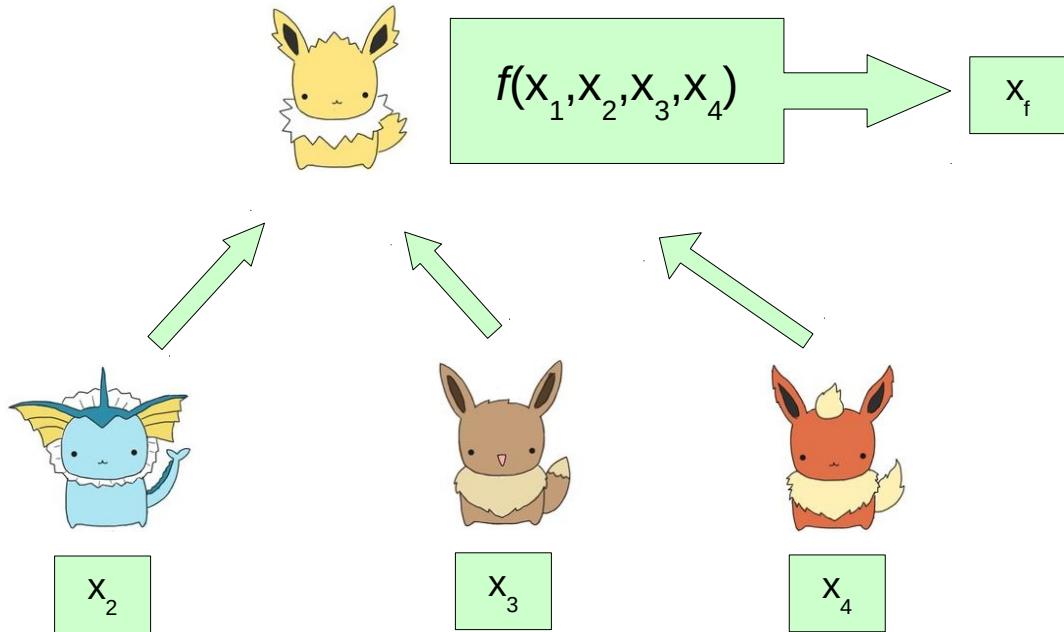
# Summary

- Extended Verifiable Computation to multiparty setting
- Used POT, Garbled Circuits, and FHE
- Compute Garbled Circuit inside FHE
- Use POT so clients  $P_{2\dots n}$  can pick from inputs sent by  $P_1$
- Next: Secure Computation

# **On-the-Fly Multiparty Computation on the Cloud via Multikey Fully Homomorphic Encryption**

# Multiparty Computation

**Without input secrecy**

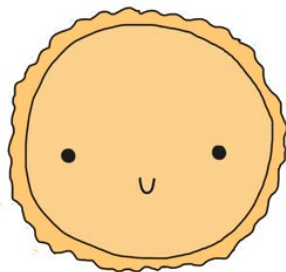




# Multiparty Computation

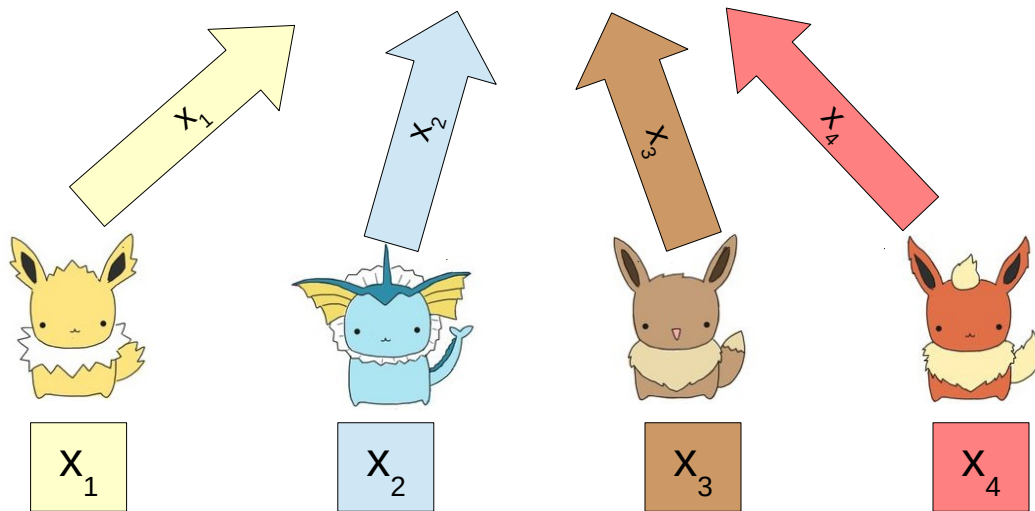
**With input secrecy**

Trusted Third Party

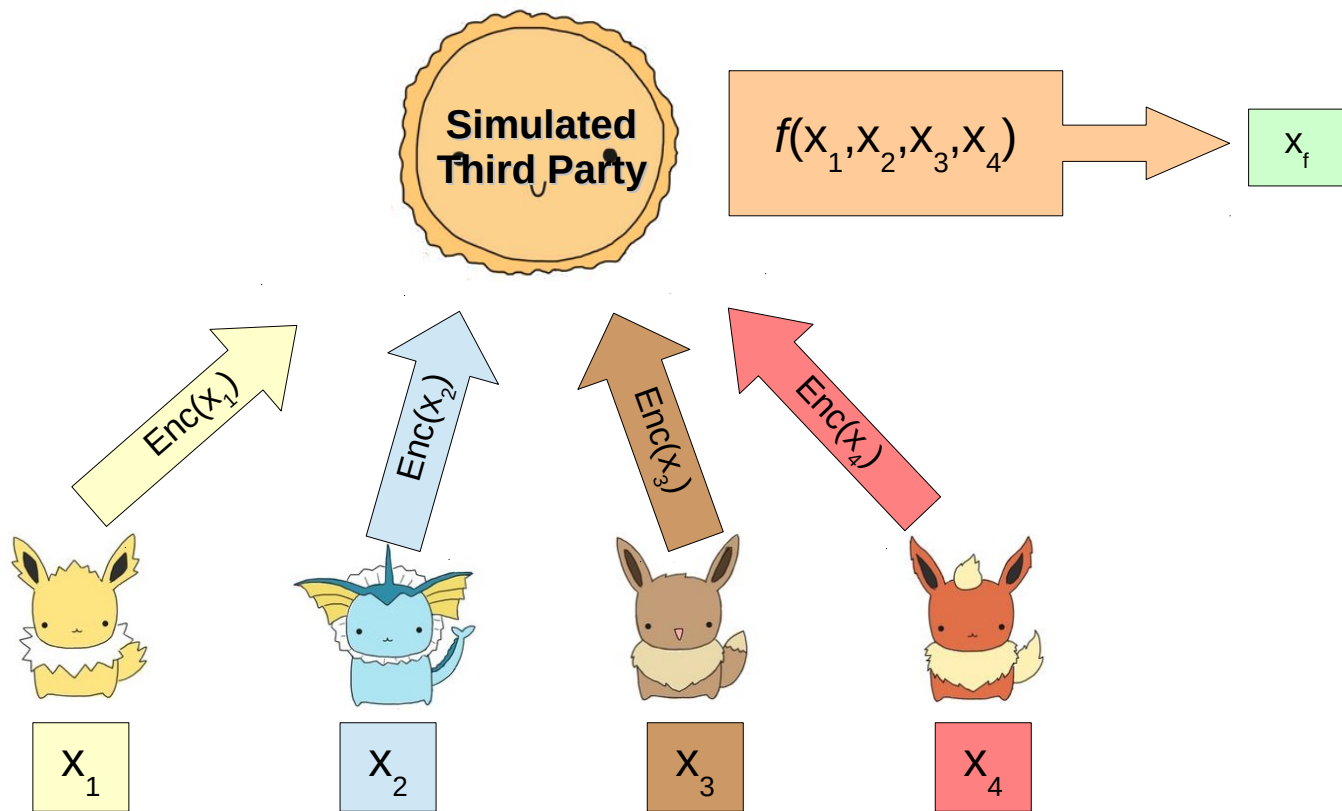


$$f(x_1, x_2, x_3, x_4)$$

$x_f$



# Secure Multiparty Computation (MPC)



# Secure Multiparty Computation (MPC)

- **Security Models**

- **Semi-Honest Adversaries** follow the protocol as described but try to learn information about honest player's input
- **Malicious Adversaries** can deviate from the protocol to learn honest player's input

# Secure Multiparty Computation (MPC)

- **Security Models**

- **Semi-Malicious Adversaries** are like semi-honest adversaries but can sample random elements from any arbitrary distribution

# Secure Multiparty Computation (MPC)

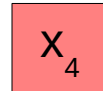
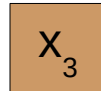
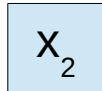
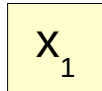
- **Problem with regular MPC?**
  - Highly interactive
  - One-time use
  - Assume all parties are equally powerful

# On-the-Fly MPC on the Cloud

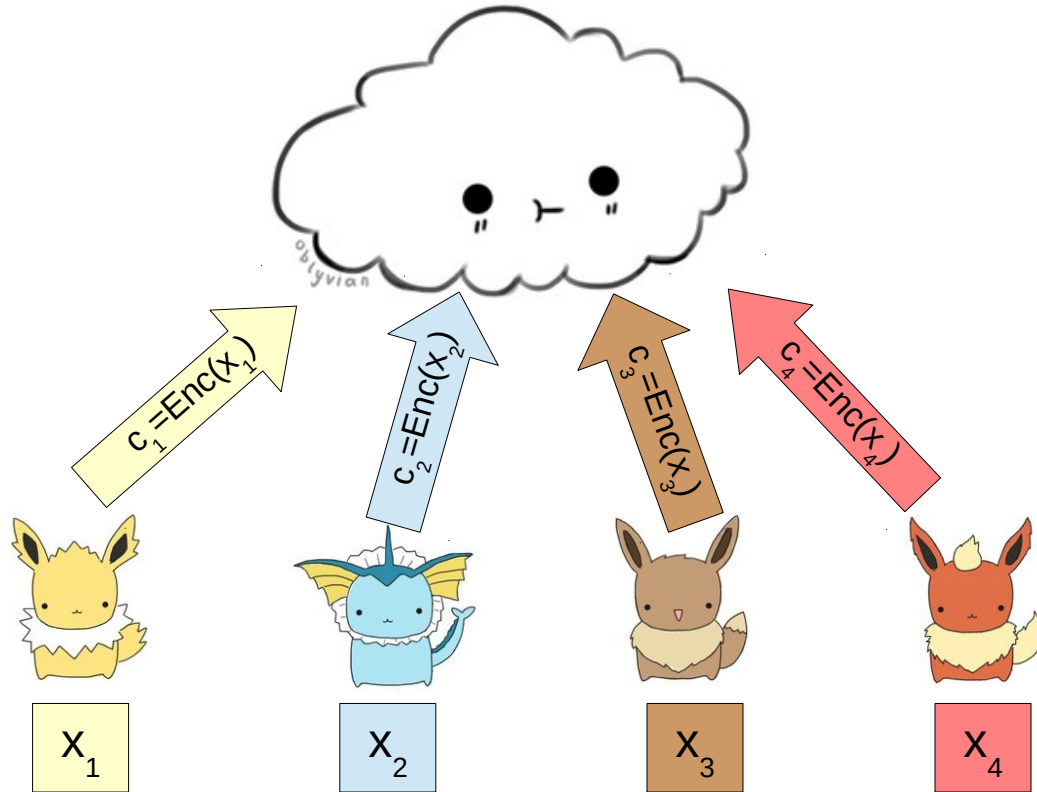
Powerful Cloud



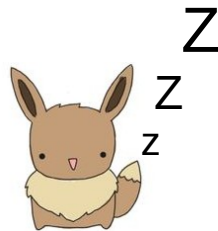
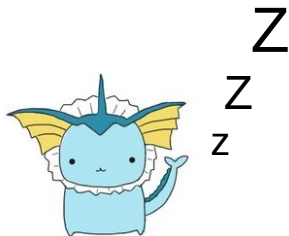
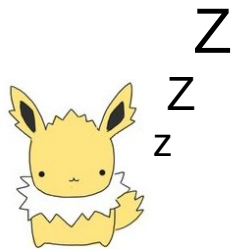
Weak Clients



# On-the-Fly MPC on the Cloud

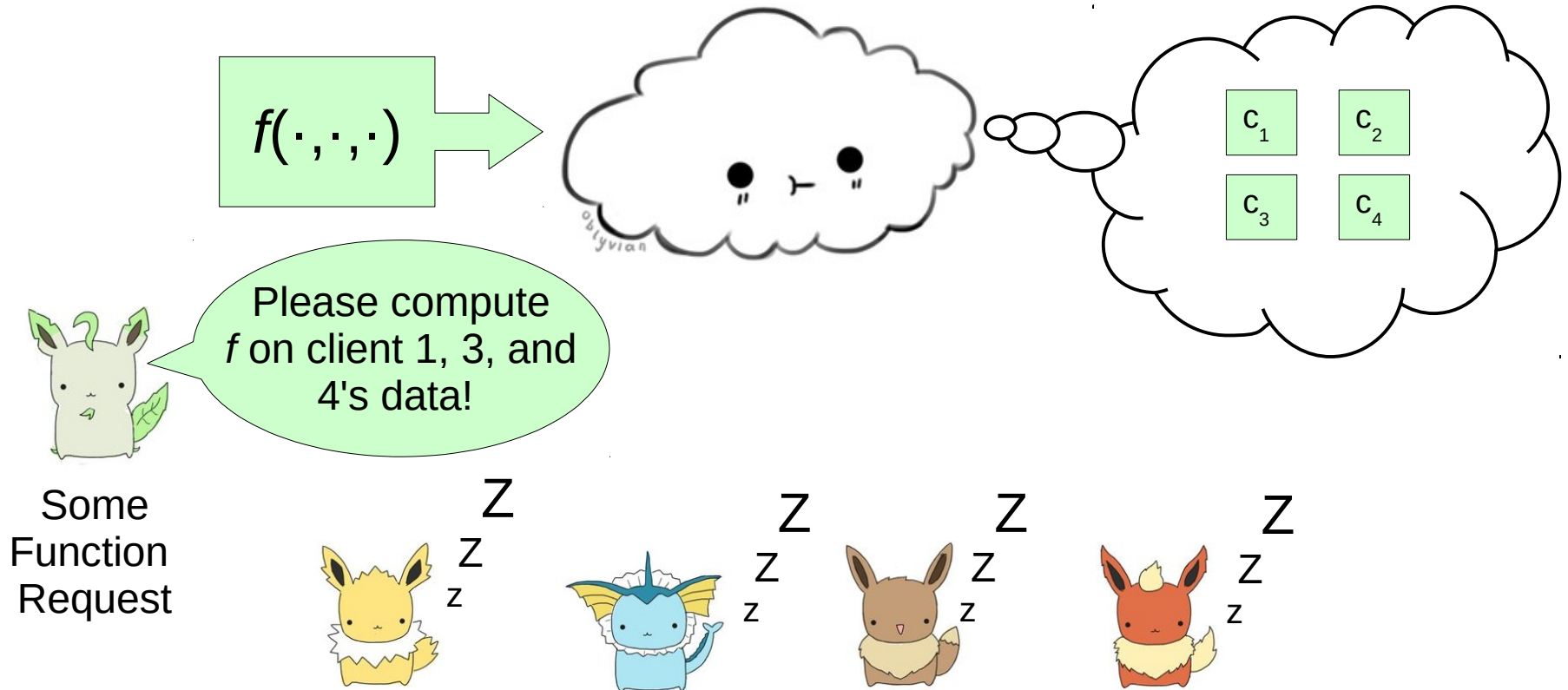


# On-the-Fly MPC on the Cloud

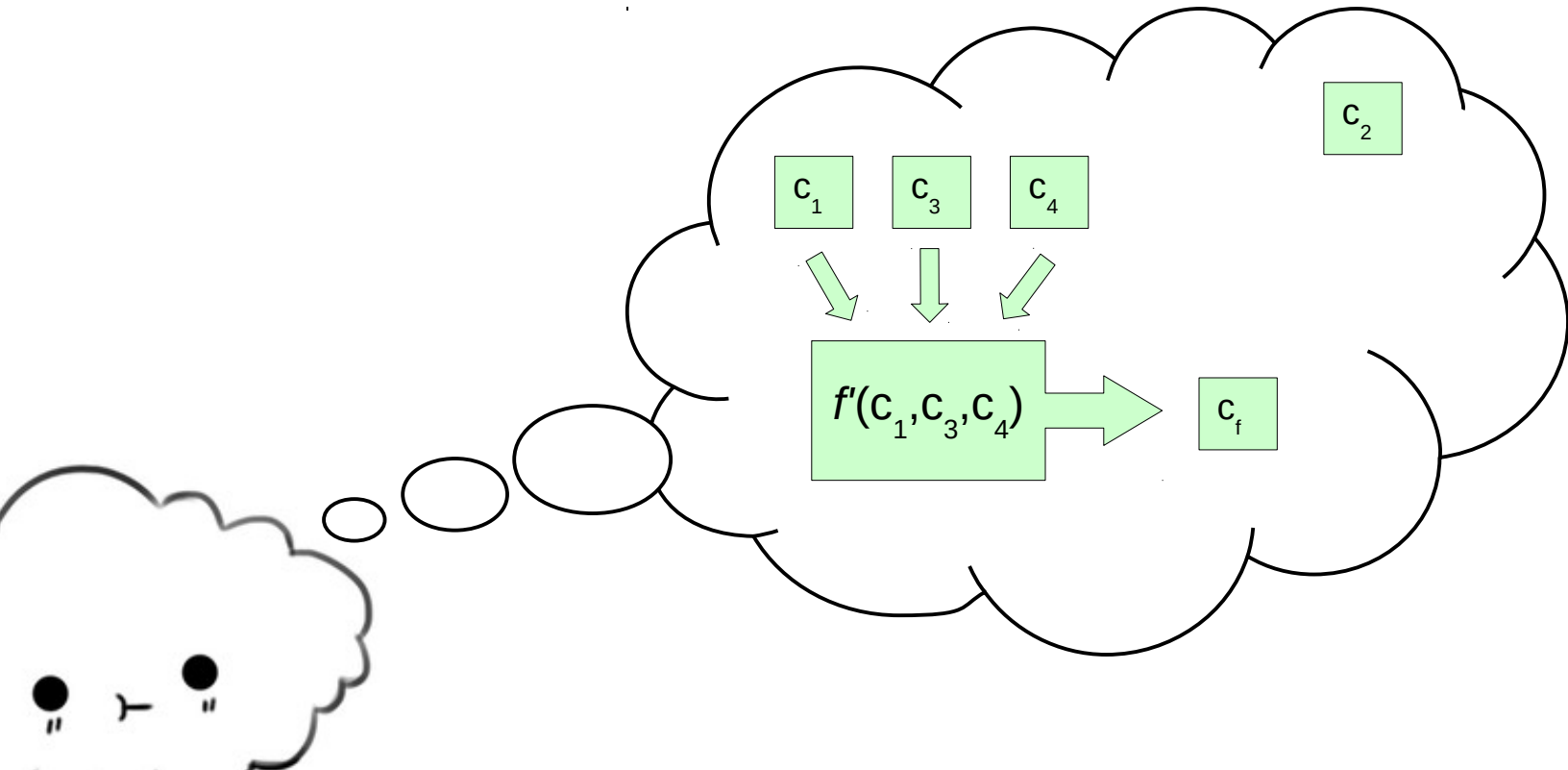




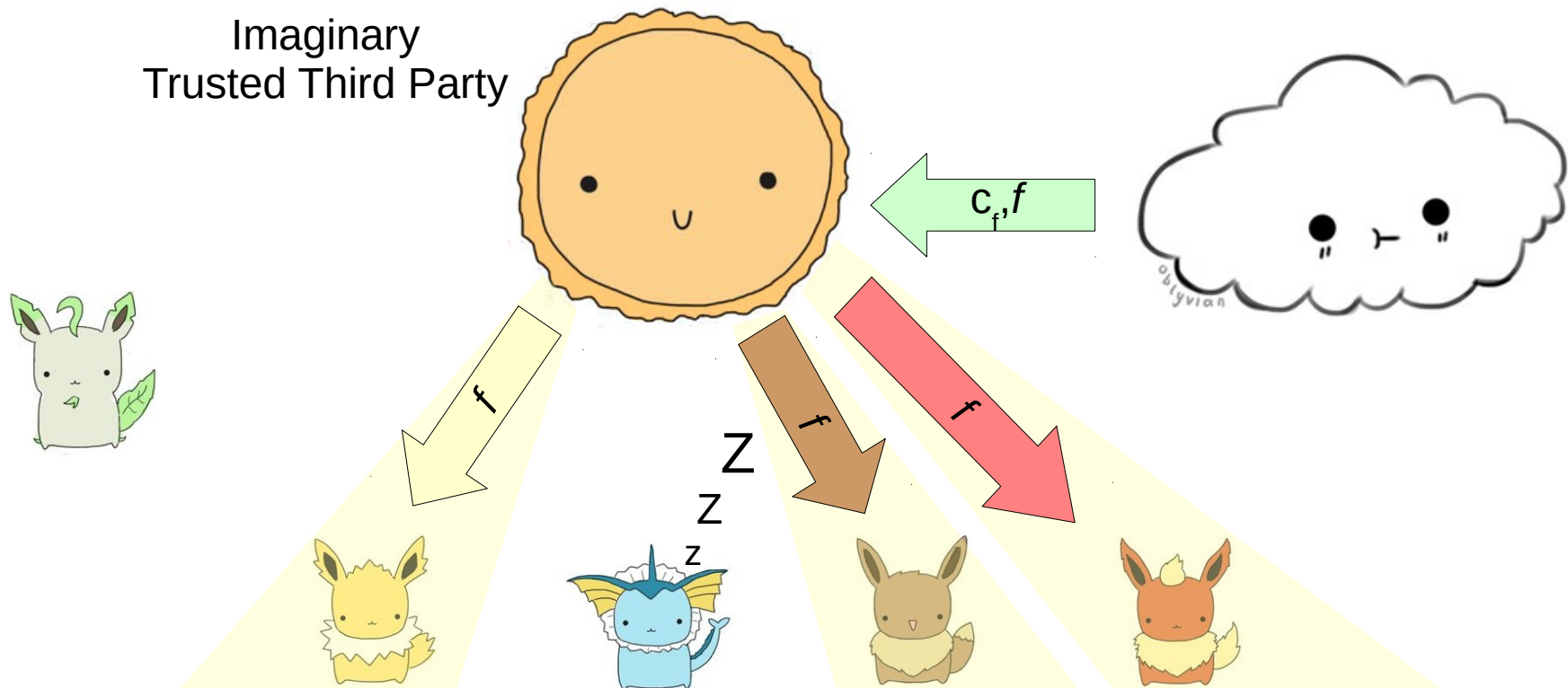
# On-the-Fly MPC on the Cloud



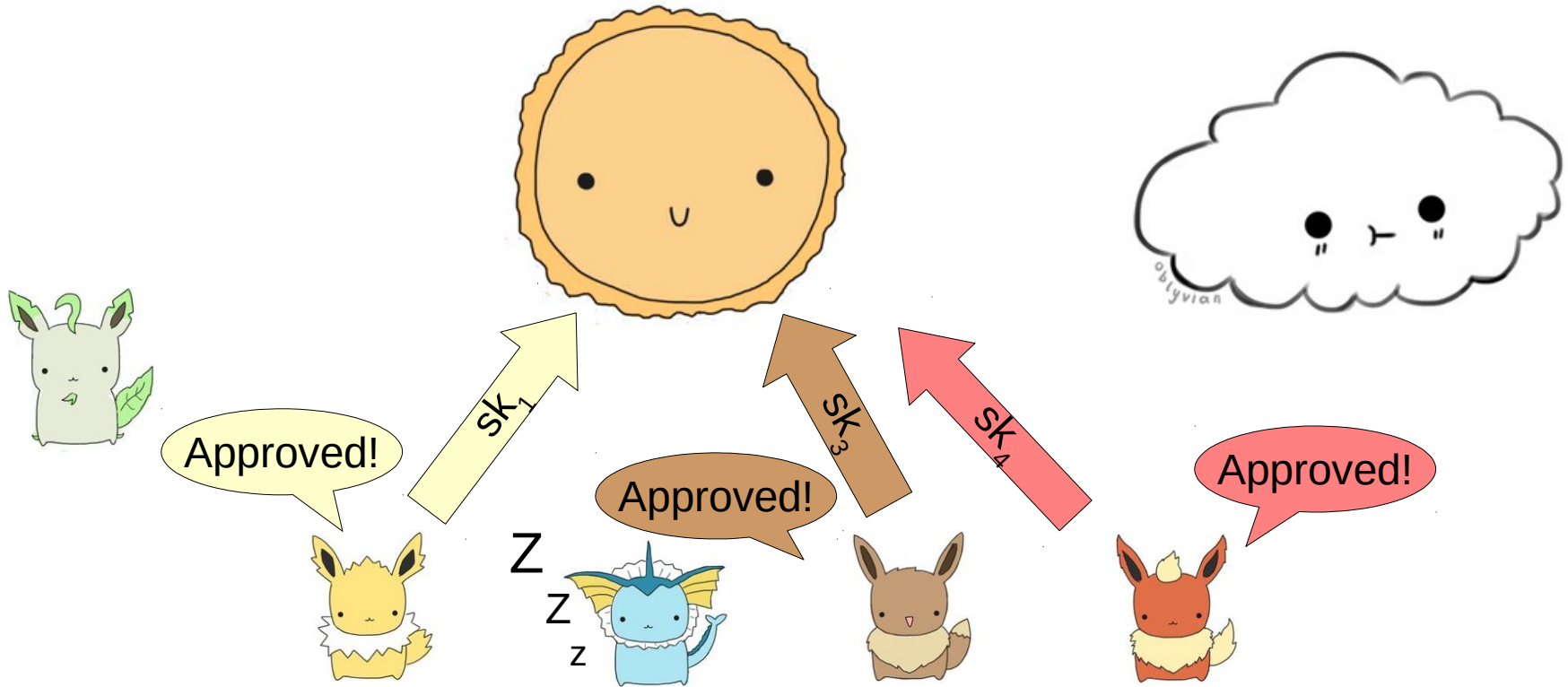
# On-the-Fly MPC on the Cloud



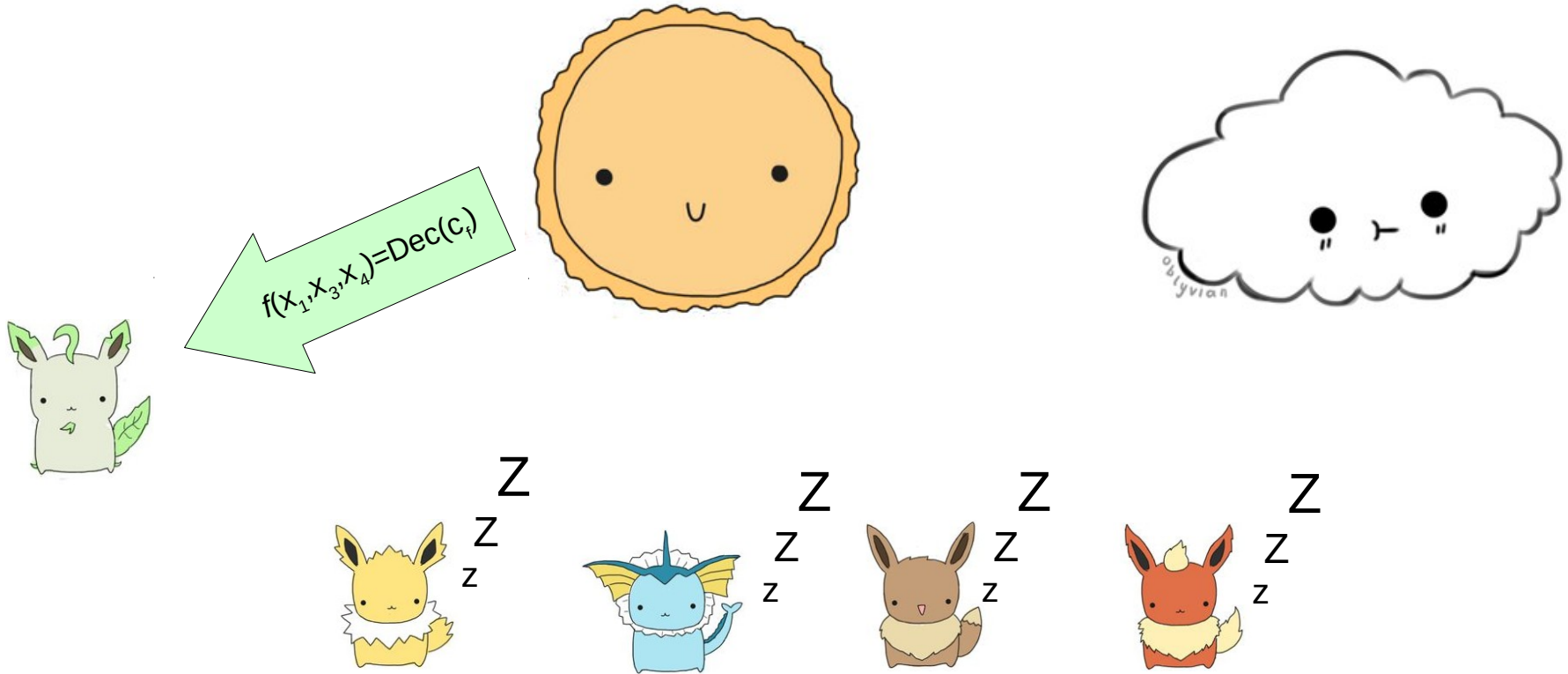
# On-the-Fly MPC on the Cloud



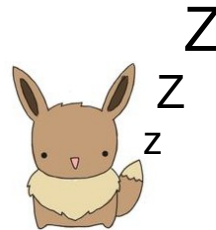
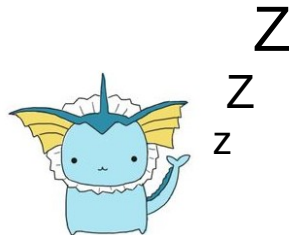
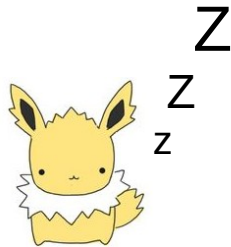
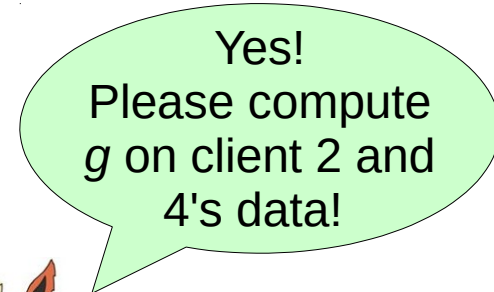
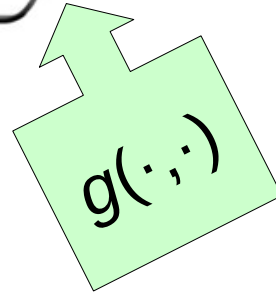
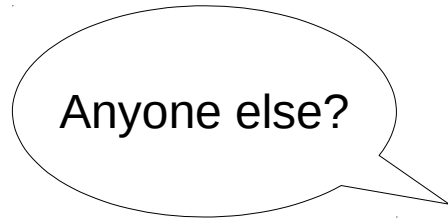
# On-the-Fly MPC on the Cloud



# On-the-Fly MPC on the Cloud



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# Properties

## **We want:**

- *Dynamically* chosen functions *on-the-fly*
- Functions compute on an *arbitrary* subset of data
- *Non-interactive* function evaluation *on the cloud*

# Properties

## **We want:**

- Clients are *unaware* of the identity or the number of other clients
- Relevant clients approve the choice of functions after evaluation and before decryption



# Semi-Malicious Construction: Building Blocks

## **Multikey Fully Homomorphic Encryption**

$$\{\mathcal{E}^{(N)} = (\text{Keygen}, \text{Enc}, \text{Dec}, \text{Eval})\}_{N \geq 0}$$

where  $N$  is the number of key pairs

- **Key Generation**

- $(pk, sk, ek) \leftarrow \text{Keygen}(1^k)$

# Semi-Malicious Construction: Building Blocks

## Multikey Fully Homomorphic Encryption

- **Encryption**

- $c \leftarrow \text{Enc}(\text{pk}, m)$

- **Decryption**

- $m := \text{Dec}(\text{sk}_1, \dots, \text{sk}_N, c)$

# Semi-Malicious Construction: Building Blocks

## Multikey Fully Homomorphic Encryption

### – Evaluation

- $c := \text{Eval}(C, (c_1, pk_1, ek_1), \dots, (c_l, pk_l, ek_l))$

where each tuple of  $\{(pk_i, sk_i, ek_i)\}_{i \in [l]}$  is in

$\{(pk_j, sk_j, ek_j)\}_{j \in [N]}$

# Semi-Malicious Construction: Building Blocks

## Multikey Fully Homomorphic Encryption

- **Correctness**

- $\text{Dec}(\text{sk}_1, \dots, \text{sk}_N, c) = C(m_1, \dots, m_l)$

- **Compactness**

- $|c| \leq P(k, N)$

# Semi-Malicious Construction: Building Blocks

## Multikey Fully Homomorphic Encryption

- **Compactness** is important, since we don't want the running time of decryption to depend on the size of circuits

# Semi-Malicious Construction: Building Blocks

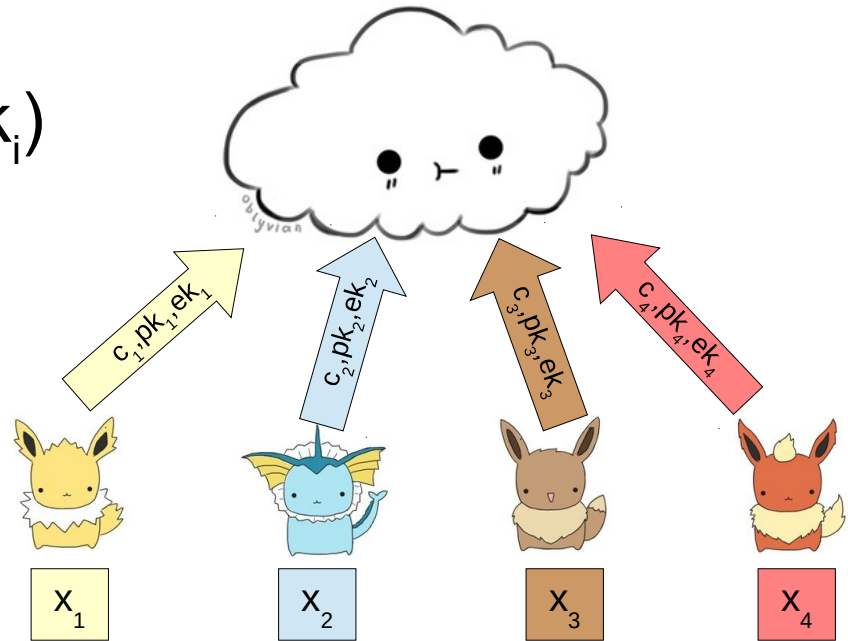
## **Semi-Malicious Secure MPC**

- **Server-Aided Secure MPC**
  - Powerful server and weak clients
  - Most computation are carried out by the server
  - Clients only communicate with the server

# Semi-Malicious Construction: Where the Building Blocks fit

## Offline Phase

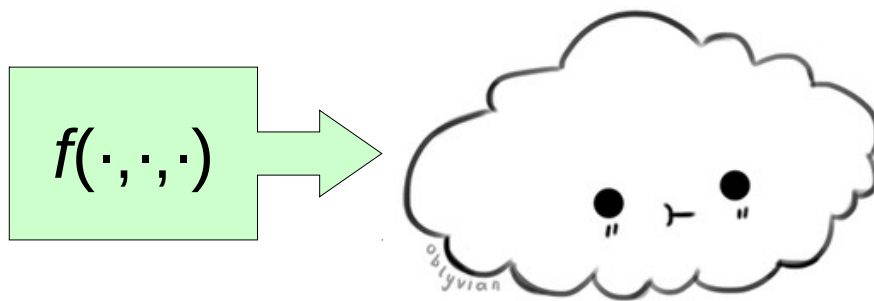
- 1) Clients samples  $(pk_i, sk_i, ek_i)$
- 2)  $c_i \leftarrow \text{Enc}(pk_i, x_i)$
- 3) Send  $c_i, pk_i, ek_i$  to the server



# Semi-Malicious Construction: Where the Building Blocks fit

## Online Phase

- Choose a function

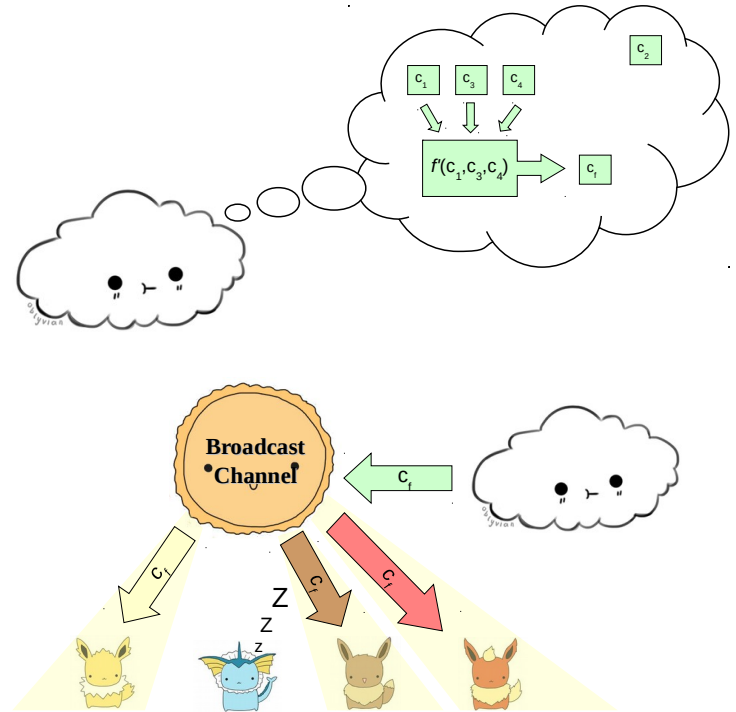




# Semi-Malicious Construction: Where the Building Blocks fit

## Online Phase: Step 1

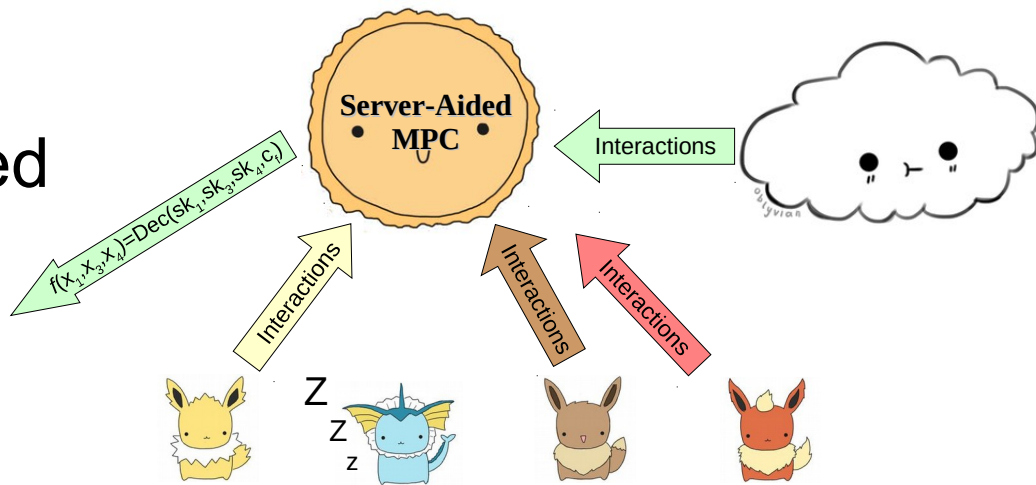
- 1) The server performs multikey FHE
- 2) The server broadcasts the evaluated ciphertext to all computing parties



# Semi-Malicious Construction: Where the Building Blocks fit

## Online Phase: Step 2

- Run server-aided MPC protocol to decrypt the evaluated ciphertext



# Problems with the Semi-Malicious Construction

- 1) Clients may encrypt data *incorrectly*
- 2) Server may compute functions *incorrectly*
- 3) Clients may cheat in semi-malicious MPC
- 4) Clients may use an invalid secret key to decrypt the resulting ciphertext

# Malicious Construction

*Clients may encrypt data incorrectly*

*Fix:* Clients must prove this NP relation to server:

$$\mathbf{R}^{\text{ENC}} = \{ ( (\mathbf{pk}, \mathbf{c}), (\mathbf{x}, \mathbf{s}) ) \mid \mathbf{c} = \text{Enc}(\mathbf{pk}, \mathbf{x}; \mathbf{s}) \}$$

where  $\mathbf{s}$  is some random string

*How?* **Zero-knowledge Proof**

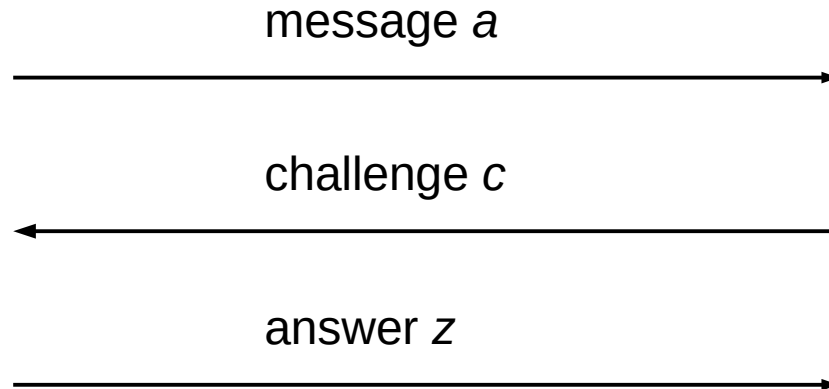
# Malicious Construction

## Non-Interactive Zero Knowledge (NIZK)

- (Interactive) Zero-Knowledge Proof

**Prover**

**Verifier**



# Malicious Construction

## Non-Interactive Zero Knowledge (NIZK)

- NIZK

**Prover**

**Verifier**

common reference string (CRS)



message  $a$ ,  
challenge  $c$ ,  
answer  $z$



# Malicious Construction

*Server may compute functions incorrectly*

*Fix:* Server verifies its computation to clients

- Use **succinct non-interactive arguments of knowledge (SNARK)**
- Similar to PCP theorem, but non-interactive

# Malicious Construction

***Clients may cheat in semi-malicious MPC***

*Fix:* Replace semi-malicious MPC with malicious MPC



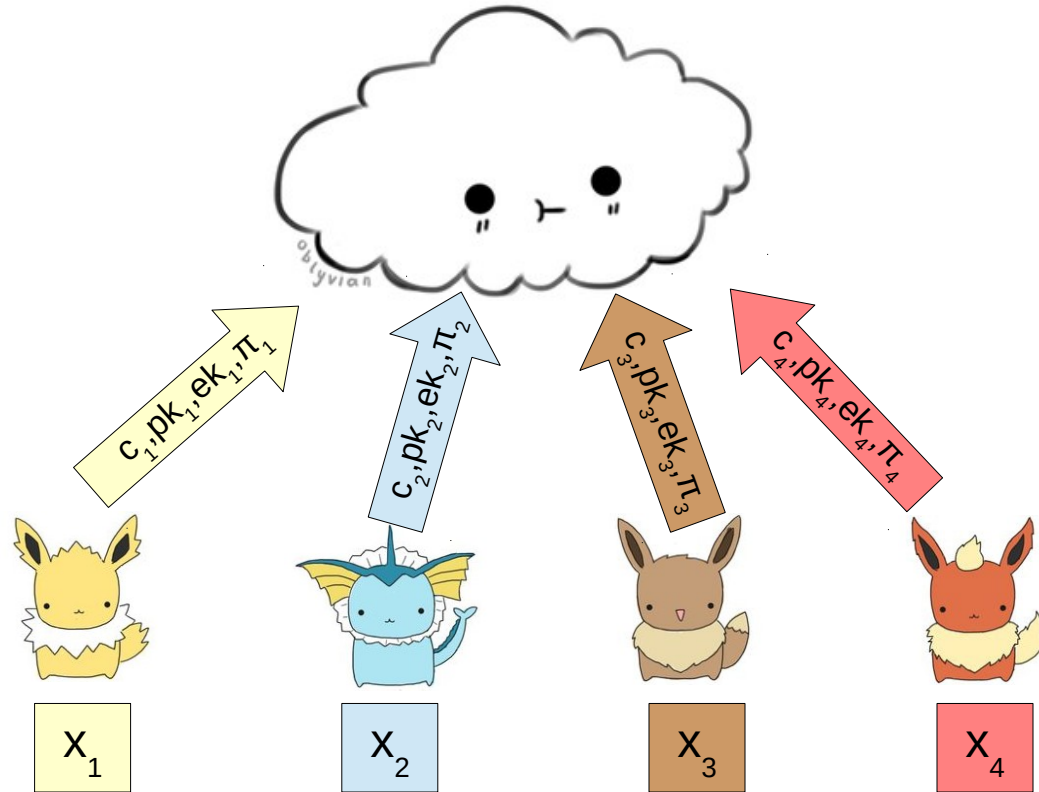
# Malicious Construction

***Clients may use an invalid secret key to decrypt the resulting ciphertext***

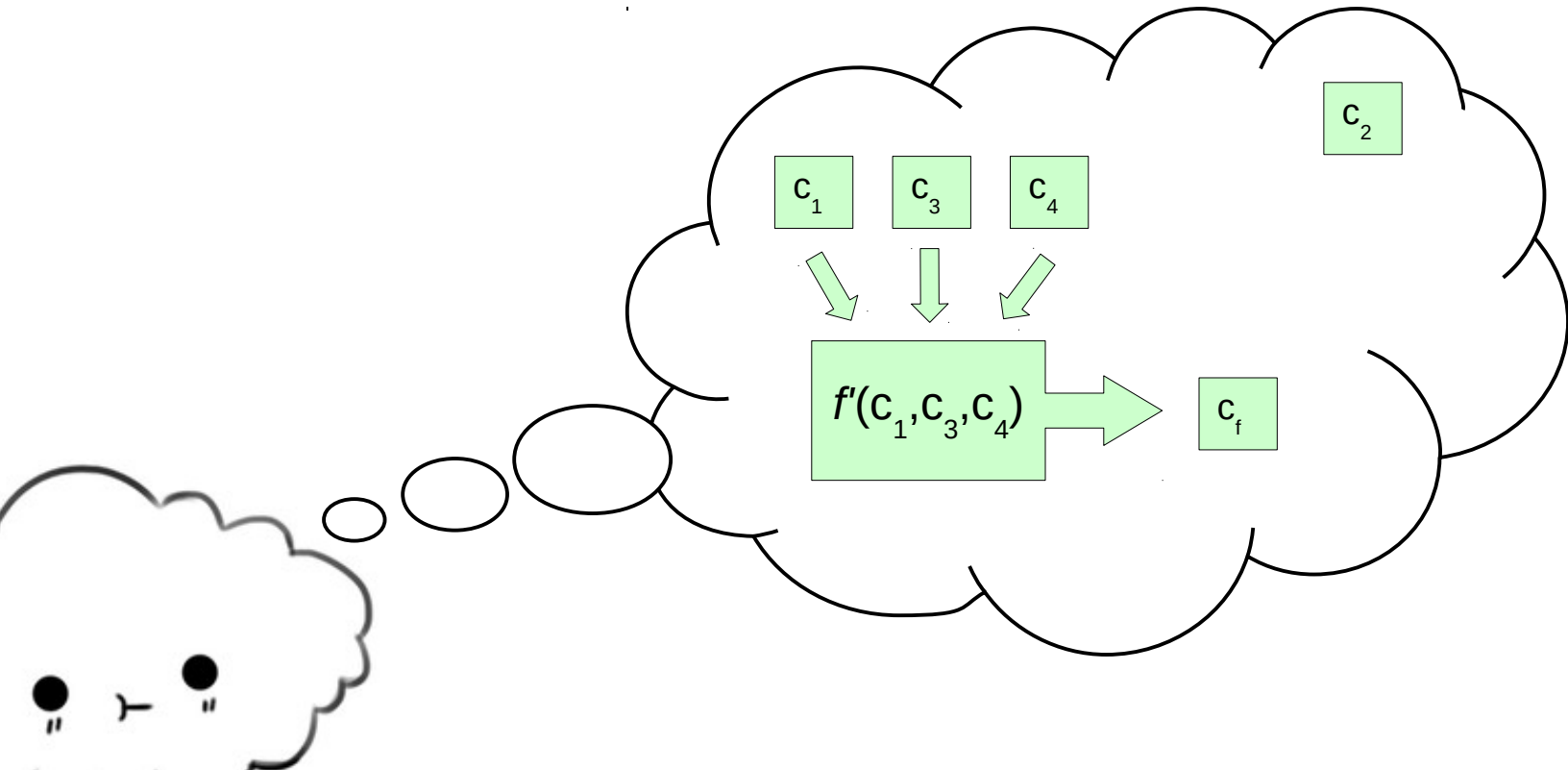
*Fix:* Change the decryption function to this:

$$g_{c, pk_1, ek_1, \dots, pk_N, ek_N}((sk_1, r_1) \dots, (sk_N, r_N)) \\ \stackrel{\text{def}}{=} \begin{cases} \text{Dec}(sk_1, \dots, sk_N, c) & \text{if } (pk_i, sk_i, ek_i) = \text{Keygen}(1^\kappa; r_i) \quad \forall i \in [N] \\ \perp & \text{otherwise} \end{cases}$$

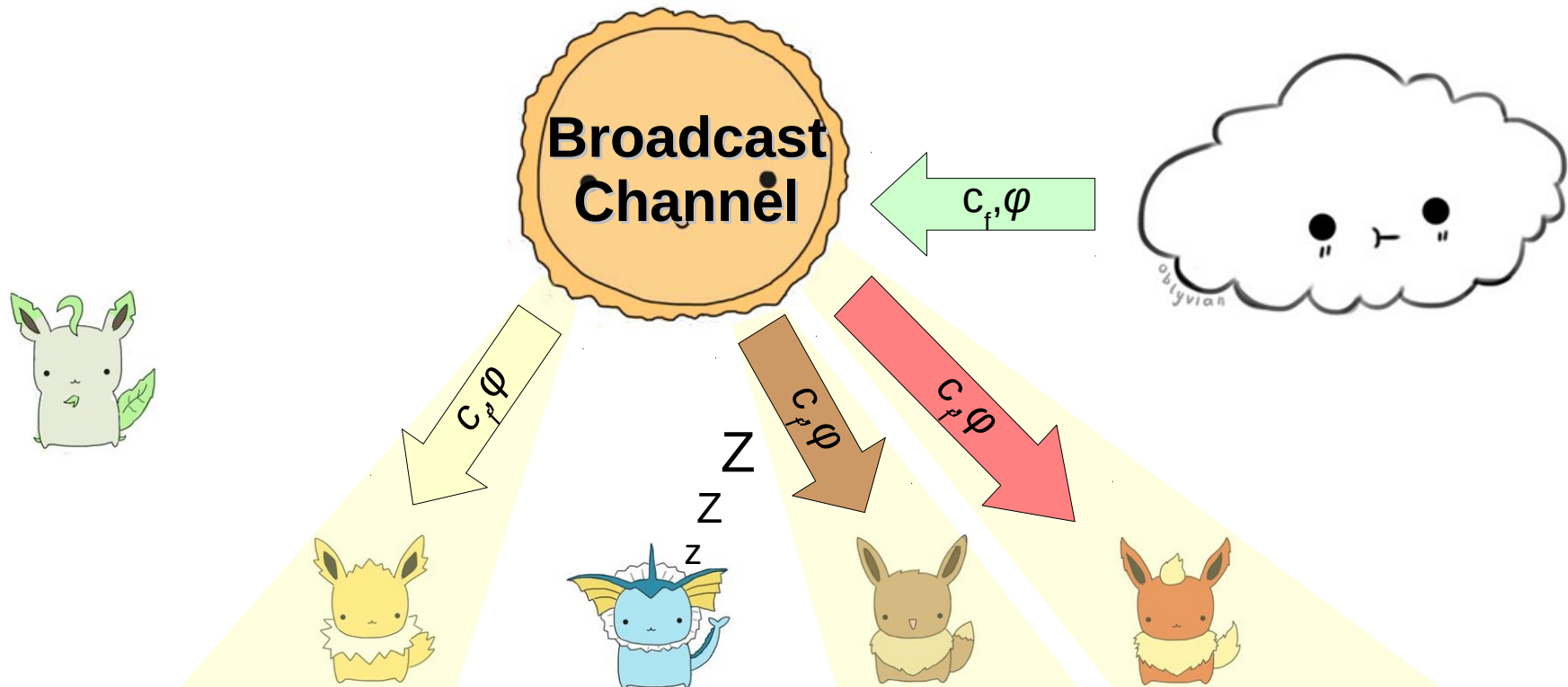
# Malicious Construction: Summary



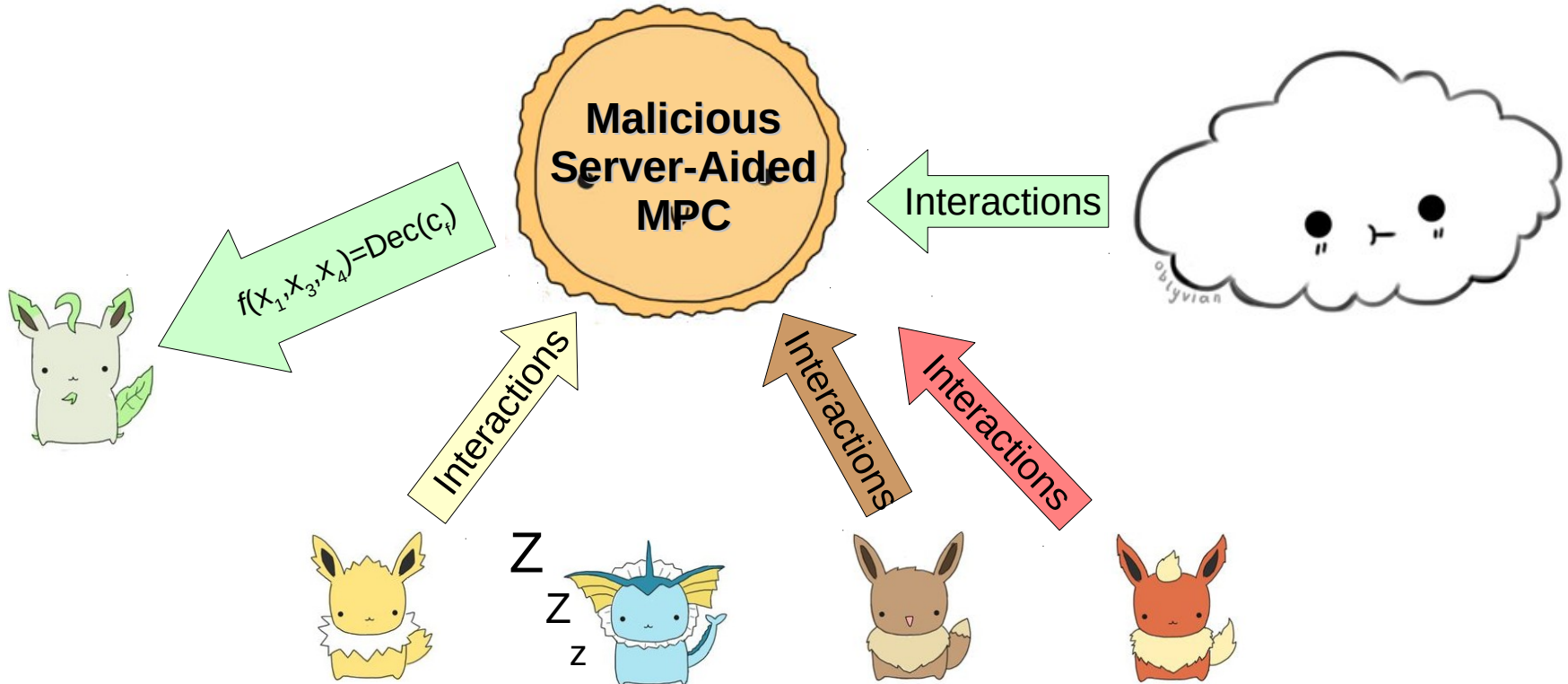
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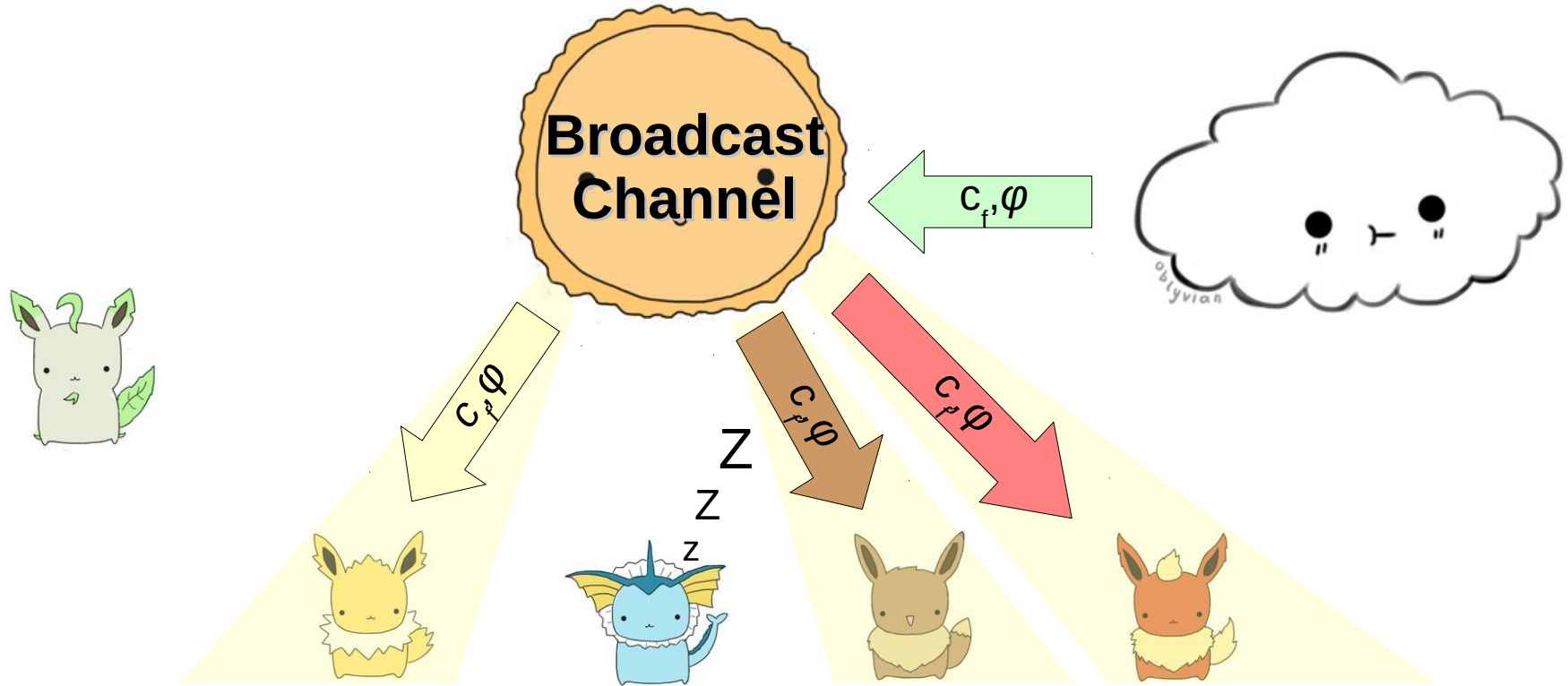


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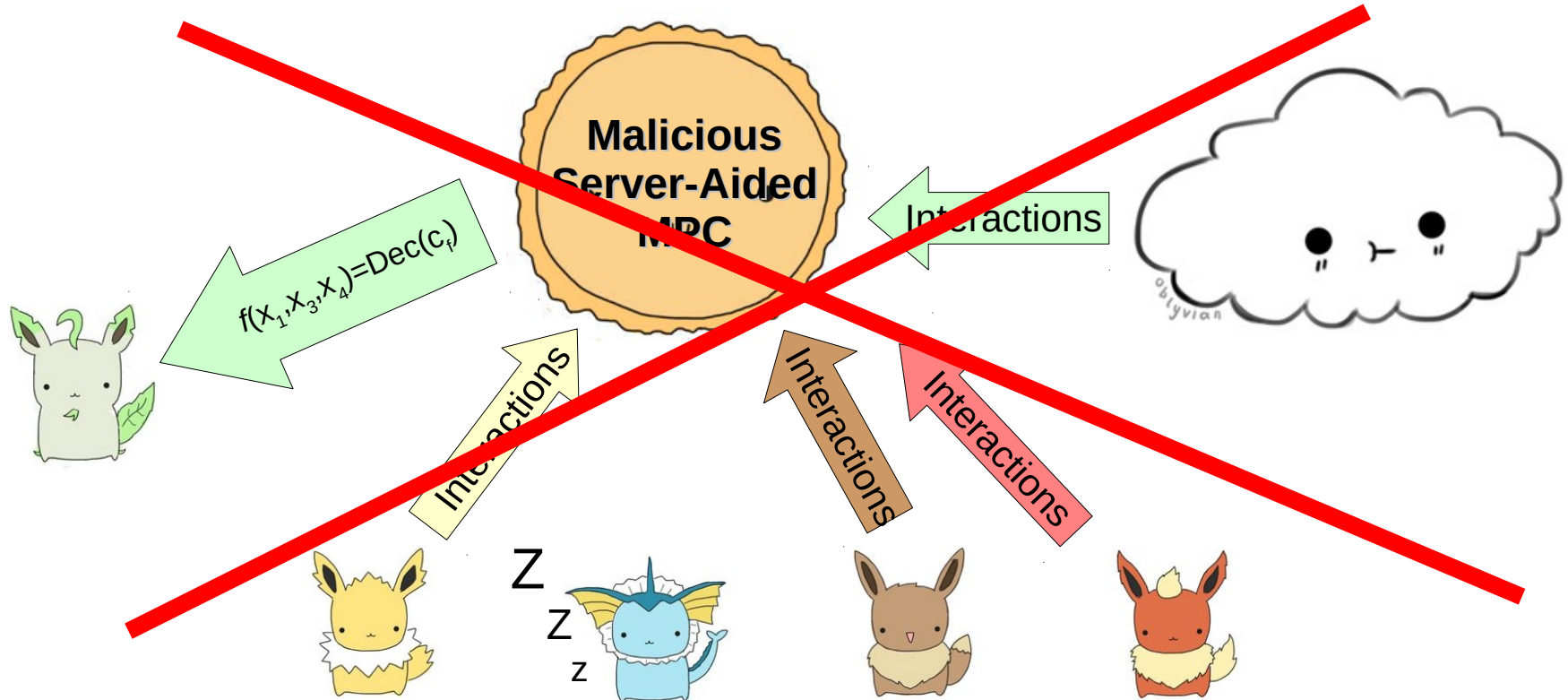


**How about a protocol with  
fewer interactions?**

# Protocol with Fewer Interactions

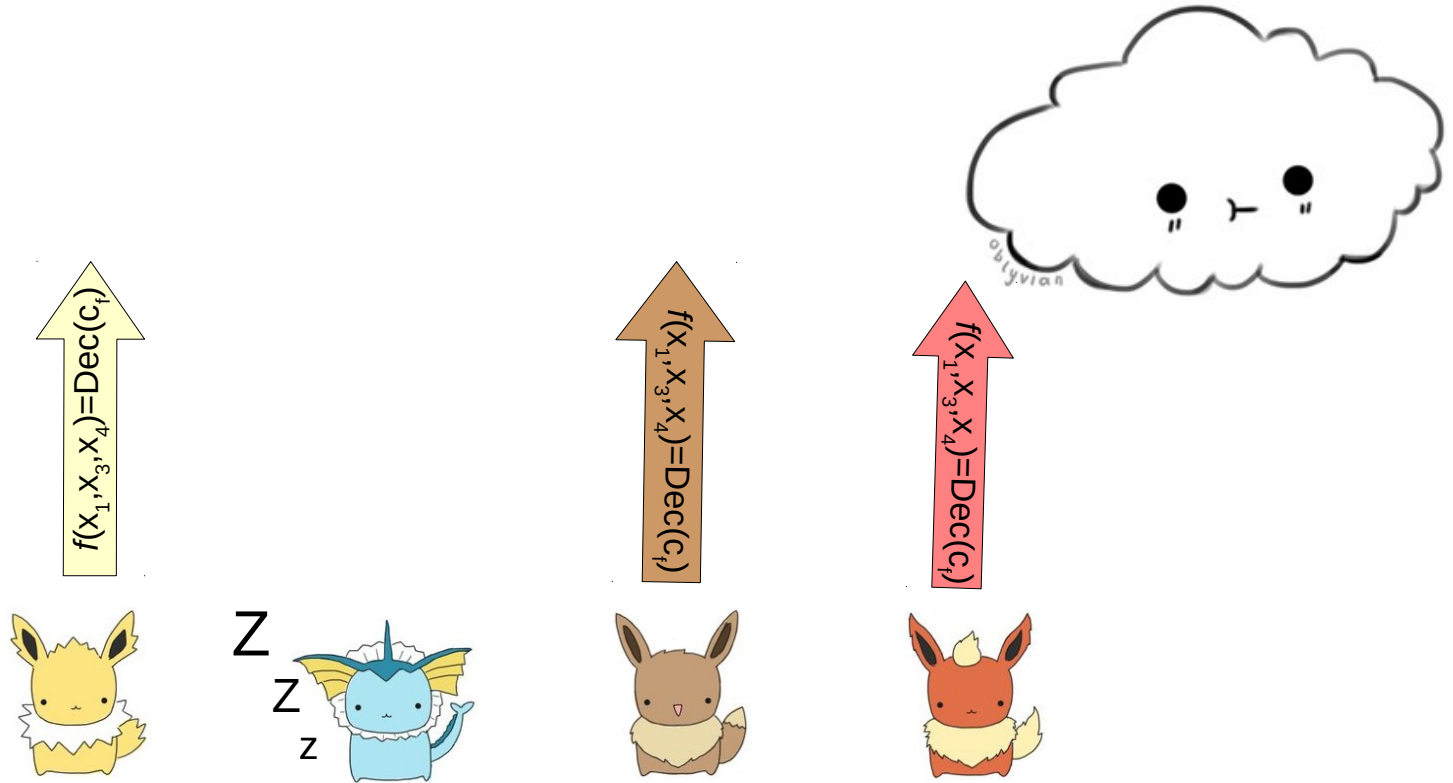


# Protocol with Fewer Interactions

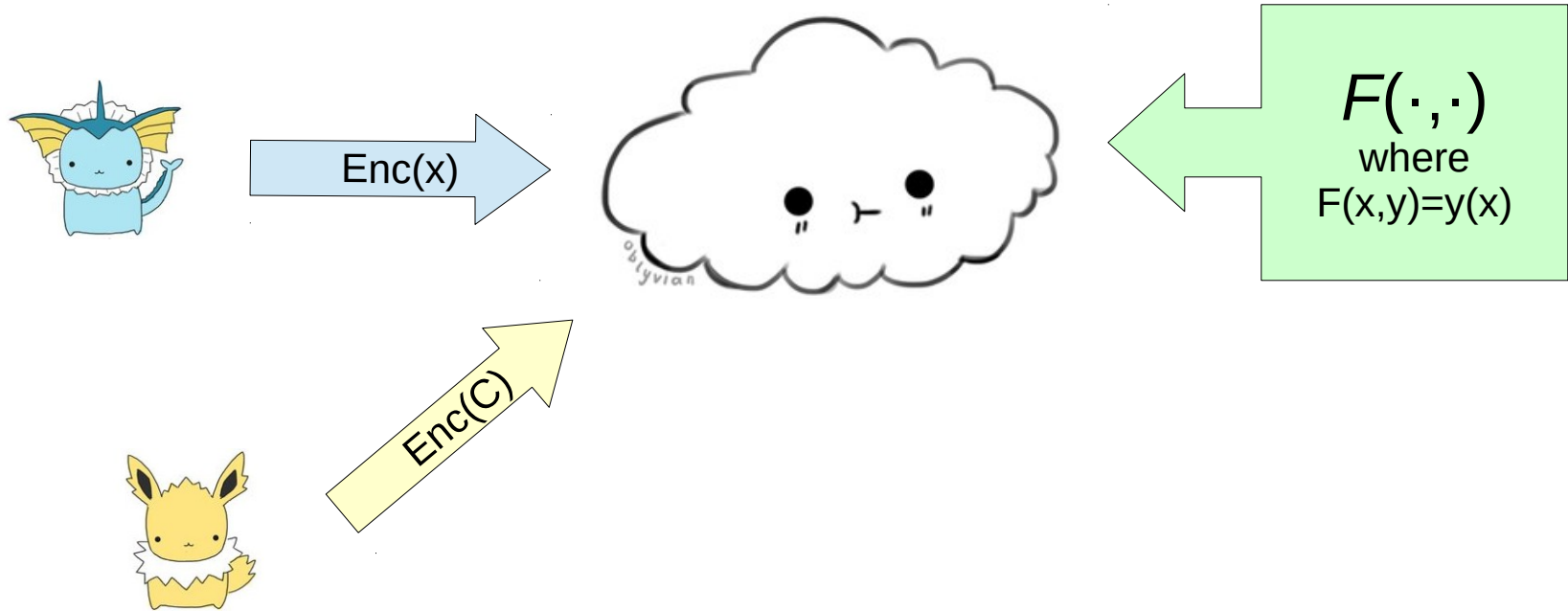




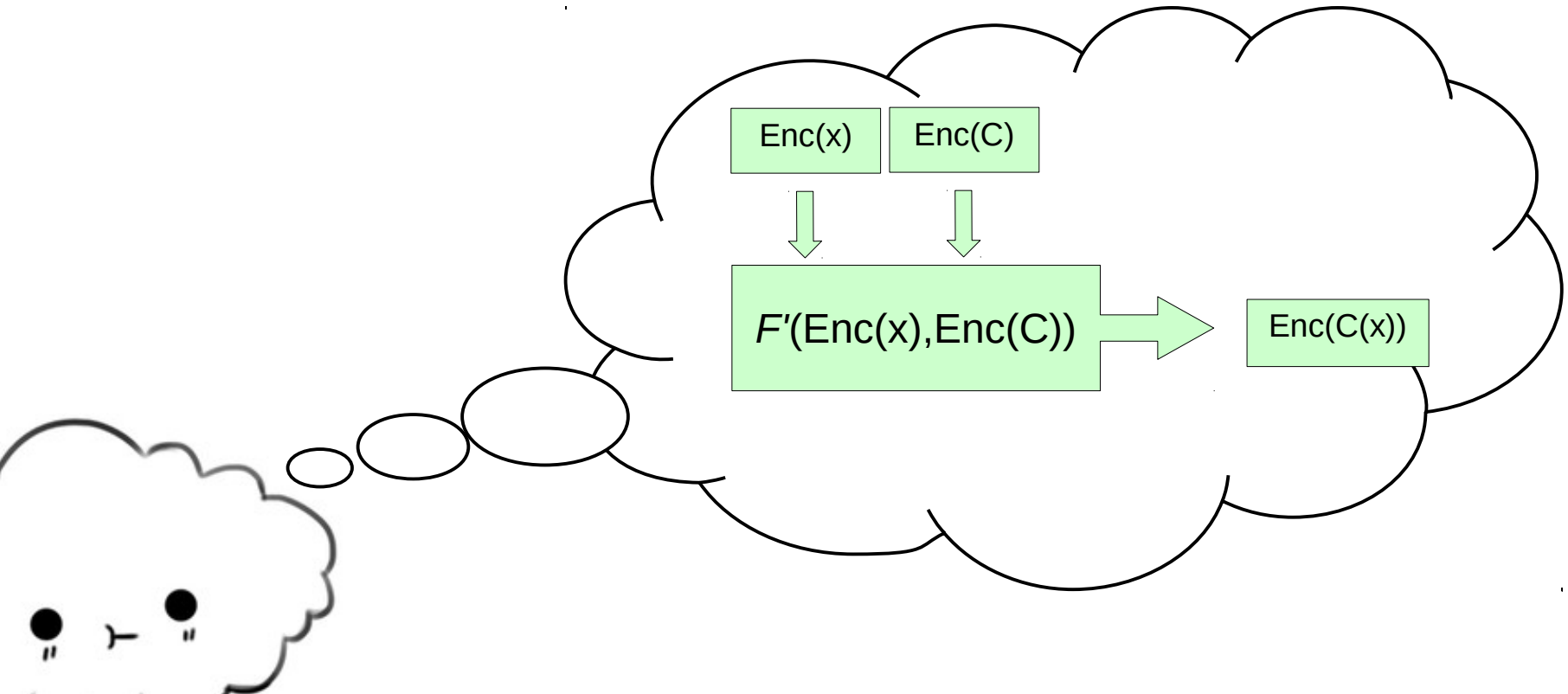
# Protocol with Fewer Interactions



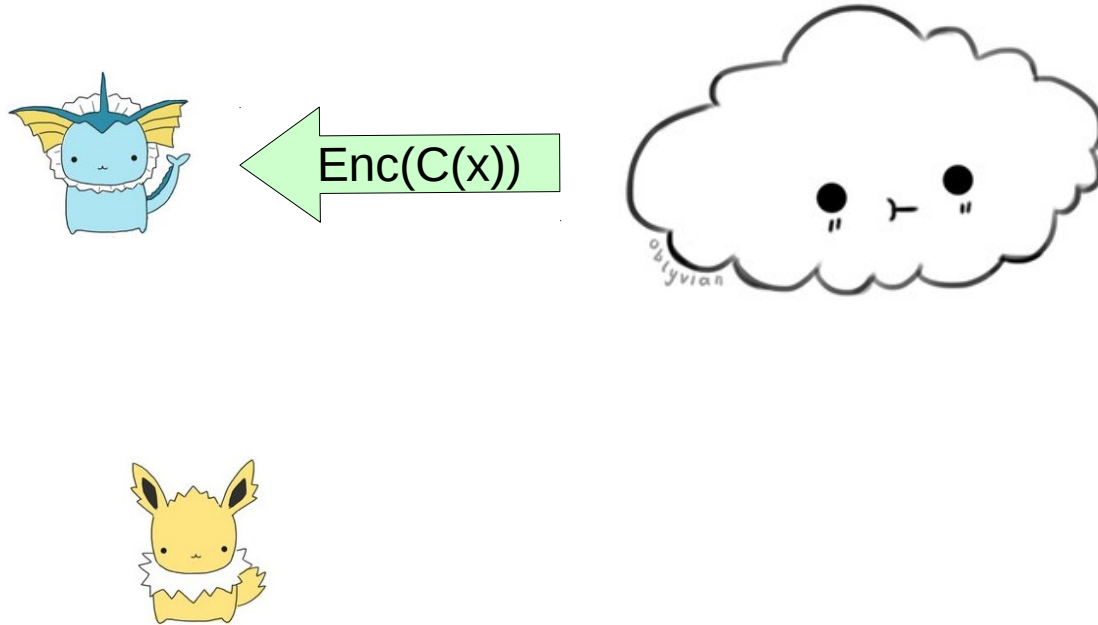
# Reduction to VBB Obfuscation



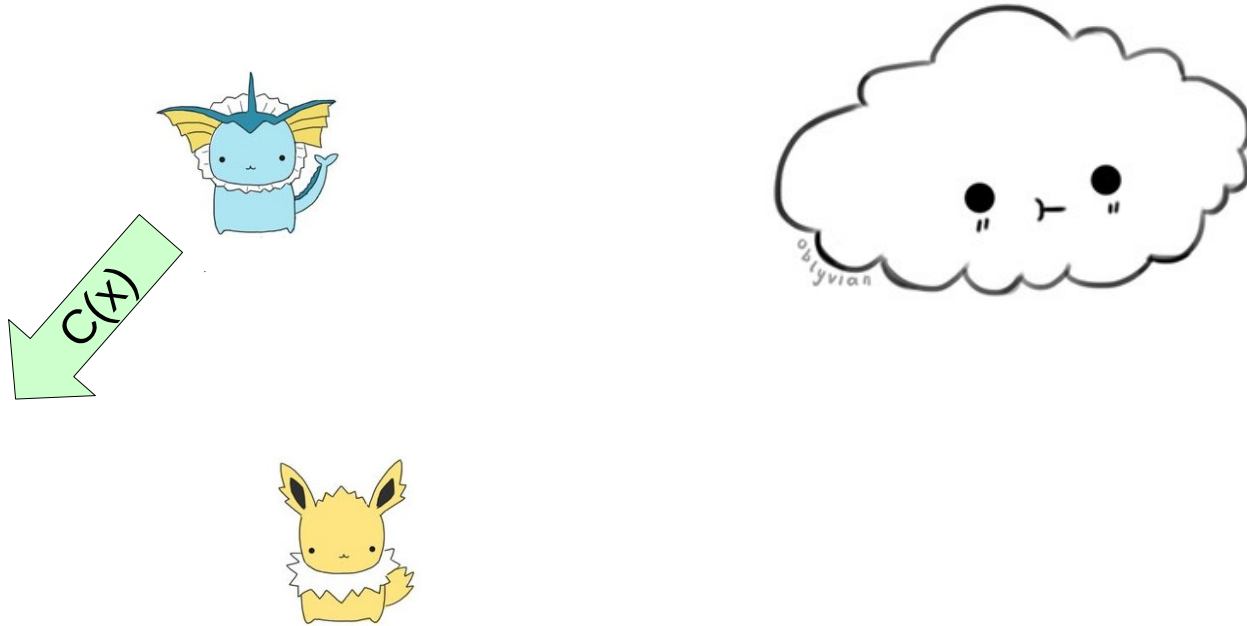
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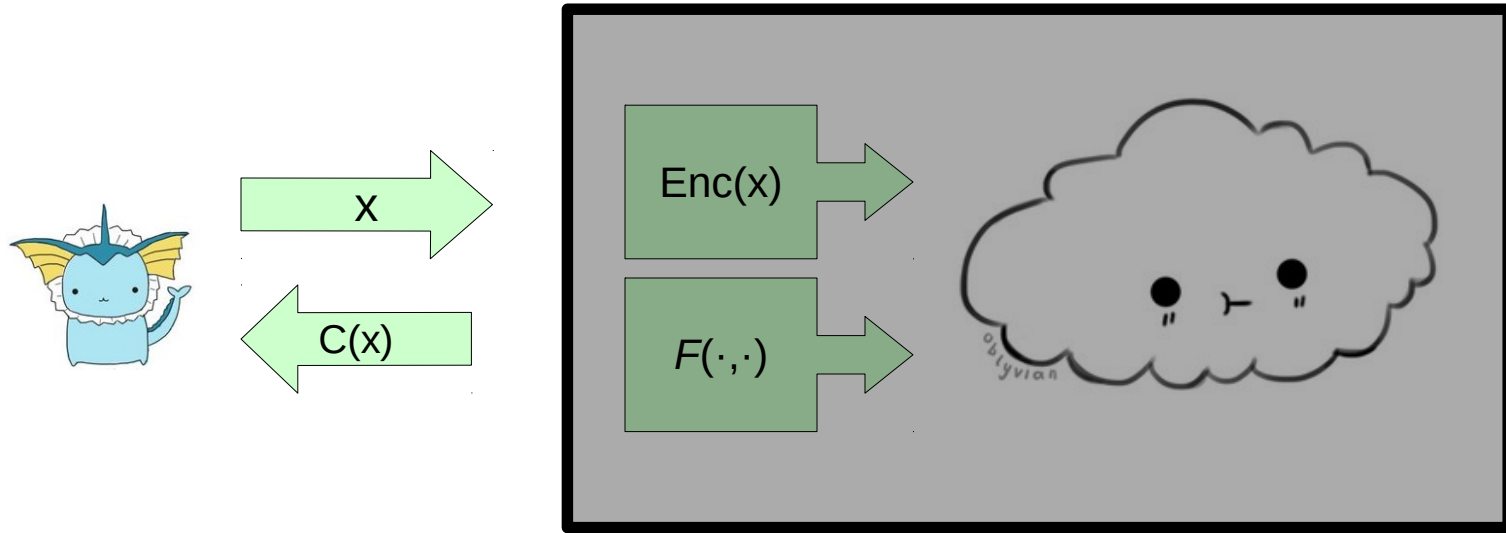
# Reduction to VBB Obfuscation



# Reduction to VBB Obfuscation



# Reduction to VBB Obfuscation



**Questions?**