

# Task 1: Secure Multiparty Computation using Beaver Triples

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## 1. Introduction

This document explains the concept, implementation, and detailed working of **Secure Multiparty Computation (MPC)** using **Beaver Triples**. The project simulates a 3-party computation that multiplies secret values without revealing the inputs to any of the participating parties.

## 2. What is MPC?

**Secure Multiparty Computation (MPC)** allows a group of parties to compute a function over their inputs while keeping those inputs private. No individual party learns anything about others' private inputs, yet the final output is correct.

**Example use cases:**

- Privacy-preserving statistics (e.g., average salary)
- Secure voting and auctions
- Federated learning and ML on private data

## 3. Key Technique: Beaver Triples

Beaver triples are preprocessed random values that enable secure multiplication:

- Random values:  $a, b$
- Their product:  $c = a \times b$
- Secret shares of  $a, b, c$  are distributed to each party

**Secure Multiplication Formula:**

$$x \cdot y = d \cdot e + d \cdot b + e \cdot a + c$$

Where:

- $d = x - a$
- $e = y - b$

This trick allows parties to securely compute  $x \cdot y$  without revealing  $x$  or  $y$ .

## 4. Project Setup and Flow

### 4.1. Components

- **coordinator.py** – acts as server; distributes shares, orchestrates protocol, reconstructs result
- **party.py** – each party receives shares, performs secure local computation, and sends result back

### 4.2. Workflow

1. The coordinator secret-shares inputs  $x$  and  $y$  among 3 parties
2. A Beaver triple  $(a, b, c = ab)$  is generated and also shared
3. Each party computes:

$$d_i = x_i - a_i, \quad e_i = y_i - b_i$$

4. The coordinator collects all  $d_i, e_i$  and computes:

$$d = x - a = \sum d_i, \quad e = y - b = \sum e_i$$

5. The coordinator computes  $d \cdot e$ , shares it again among the parties
6. Each party computes:

$$\text{share}_i = \text{de\_share}_i + d \cdot b_i + e \cdot a_i + c_i$$

7. The coordinator collects and sums the 3 shares to get the final result:

$$x \cdot y = \sum \text{share}_i$$

## 5. Why Coordinator Computes $d \cdot e$

Even though  $d$  and  $e$  are public, the value  $d \cdot e$  is re-shared to prevent all parties from seeing it directly. If they saw the full  $d \cdot e$ , it would leak intermediate data — violating MPC's goal.

## 7. Worked Example for Revision

**Goal:** Securely compute  $x \times y = 8 \times 5 = 40$  using 3-party MPC with Beaver triples.

### Step 1: Secret Share Inputs

We split  $x = 8$  and  $y = 5$  into 3 shares each:

$$x_1 = 2, \quad x_2 = 3, \quad x_3 = 3 \Rightarrow x = x_1 + x_2 + x_3 = 8$$

$$y_1 = 1, \quad y_2 = 2, \quad y_3 = 2 \Rightarrow y = y_1 + y_2 + y_3 = 5$$

**Step 2: Generate Beaver Triple**

Let the Beaver triple be:

$$a = 4, \quad b = 2, \quad c = a \cdot b = 8$$

Shares of  $a$ ,  $b$ ,  $c$ :

$$a_1 = 1, \quad a_2 = 2, \quad a_3 = 1 \Rightarrow a = 4$$

$$b_1 = 0, \quad b_2 = 1, \quad b_3 = 1 \Rightarrow b = 2$$

$$c_1 = 3, \quad c_2 = 2, \quad c_3 = 3 \Rightarrow c = 8$$

**Step 3: Each Party Computes  $d_i = x_i - a_i$ ,  $e_i = y_i - b_i$** 

$$d_1 = 2 - 1 = 1, \quad e_1 = 1 - 0 = 1$$

$$d_2 = 3 - 2 = 1, \quad e_2 = 2 - 1 = 1$$

$$d_3 = 3 - 1 = 2, \quad e_3 = 2 - 1 = 1$$

**Step 4: Coordinator Reconstructs**

$$d = d_1 + d_2 + d_3 = 4, \quad e = e_1 + e_2 + e_3 = 3$$

$$d \cdot e = 4 \cdot 3 = 12$$

**Step 5: Re-share  $d \cdot e = 12$** 

Let the shares be:

$$de_1 = 5, \quad de_2 = 4, \quad de_3 = 3$$

**Step 6: Each Party Computes Final Share**

$$\text{Party 1: } 5 + 4 \cdot 0 + 3 \cdot 1 + 3 = 5 + 0 + 3 + 3 = 11$$

$$\text{Party 2: } 4 + 4 \cdot 1 + 3 \cdot 2 + 2 = 4 + 4 + 6 + 2 = 16$$

$$\text{Party 3: } 3 + 4 \cdot 1 + 3 \cdot 1 + 3 = 3 + 4 + 3 + 3 = 13$$

**Step 7: Coordinator Reconstructs Final Result**

$$\text{Result} = 11 + 16 + 13 = 40 \quad (\text{Correct!})$$

This confirms the secure computation of  $x \cdot y$  using 3-party MPC and Beaver triples.

**6. Final Notes**

This system securely computes multiplication without revealing any inputs to individual parties. It can be extended to support:

- Dot product over vectors
- XOR-based secret sharing
- Parallel secure multiplications

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