# Secure Multiparty Computation 

and Team Matching

## Shamir Secret Sharing

Secret values a, $b: \quad a, b \in \mathbb{F}_{p}$

## Shamir Secret Sharing

Secret values a, $b: \quad a, b \in \mathbb{F}_{p}$

To share $a, b$ create two random degree $t$ polynomials $A(), B()$ as follows:

$$
A(x)=\sum_{i=0}^{t} \alpha_{i} x^{i}, \quad B(x)=\sum_{i=0}^{t} \beta_{i} x^{i}
$$

Set $A(0)=a$ and $B(0)=b$

## Shamir Secret Sharing

Secret values $a, b: \quad a, b \in \mathbb{F}_{p}$

To share $a, b$ create two random degree $t$ polynomials $A(), B()$ as follows:

$$
A(x)=\sum_{i=0}^{t} \alpha_{i} x^{i}, \quad B(x)=\sum_{i=0}^{t} \beta_{i} x^{i}
$$

Set $A(0)=a$ and $B(0)=b$
$n$ shares are then computed as: $\forall i \in\{1, \ldots, n\}: a_{i}=A(i), \quad b_{i}=B(i)$
$\left.A_{( }\right)$and $B()$ are degree $t$ polynomials and can be uniquely defined by $n=t+1$ points

## Uses:

The obvious: sharing of secrets
The secret value could be a key that is tied to \{a bitcoin wallet, disk encryption, etc\}
Recovering the key then requires cooperation!

## Uses:

The obvious: sharing of secrets
The secret value could be a key that is tied to \{a bitcoin wallet, disk encryption, etc\}
Recovering the key then requires cooperation!
Tiered key management:
Threshold $=1$ out of 1

## Uses:

The obvious: sharing of secrets
The secret value could be a key that is tied to \{a bitcoin wallet, disk encryption, etc\}
Recovering the key then requires cooperation!
Tiered key management:


Threshold $=1$ out of 1

Threshold $=2$ out of 3

## Uses:

The obvious: sharing of secrets
The secret value could be a key that is tied to \{a bitcoin wallet, disk encryption, etc\}
Recovering the key then requires cooperation!
Tiered key management:

Threshold = 1 out of 1

Threshold $=2$ out of 3

Threshold $=2$ out of 2

## Uses:

The obvious: sharing of secrets
The secret value could be a key that is tied to \{a bitcoin wallet, disk encryption, etc\}
Recovering the key then requires cooperation!
Tiered key management:


Threshold $=1$ out of 1

Threshold $=2$ out of 3

Threshold $=2$ out of 2

## Secure Multiparty Computation (BGW)

Alice, Bob, and Carla all have a secret value.
They want to learn the output of some function on their secret inputs, but they don't want the others to learn their input!

Canonical example of this is salary: several companies may want to learn what the average salary is for a role, but they want to keep their payroll information private.

## Secure Multiparty Computation (yay)



## Secure Multiparty Computation (yay)



Each party sends Shamir Secret Shares of their input to the other parties

## Secure Multiparty Computation (yay)



Each party sends Shamir Secret Shares of their input to the other parties

## Secure Multiparty Computation (yay)

Alice (1)<br>$a_{1}, b_{1}$,<br>$c_{1}$<br>Bob (2)<br>$a_{2}, b_{2}, c_{2}$

Now what?

## Carla (3)

$a_{3}, b_{3}$,
$c_{3}$

## Secure Multiparty Computation (yay)

Alice (1)<br>$a_{1}, b_{1}$,<br>They want to compute<br>Bob (2)<br>$c_{1}$<br>\[ \begin{gathered} f(a, b, c)=\operatorname{avg}(a, b, c)=<br>\operatorname{sum}(a, b, c) / 3 \end{gathered} \]

## Carla (3)

$$
\begin{aligned}
& a_{3}, b_{3}, \\
& c_{3}
\end{aligned}
$$

## Secure Multiparty Computation (yay)

Alice (1)<br>$a_{1}, b_{1}$,<br>How to compute<br>Bob (2)<br>$c_{1}$ $a+b+c$ ?

## Carla (3)

$$
\begin{aligned}
& a_{3}, b_{3}, \\
& c_{3}
\end{aligned}
$$

## Secure Multiparty Computation (yay)

## Alice (1)

$\operatorname{sum}(a, b, c)_{1}=a_{1}+b_{1}+c_{1}$

## Bob (2)

$\operatorname{sum}(a, b, c)_{2}=a_{2}+b_{2}+c_{2}$

## Add shares locally!

## Carla (3)

$$
\operatorname{sum}(a, b, c)_{3}=a_{3}+b_{3}+c_{3}
$$

## Secure Multiparty Computation (yay)

Alice's secret is $A(0)$ where $A()=\alpha_{0} x^{0}+\alpha_{1} x^{1}+\alpha_{2} x^{2} \quad\left(\alpha_{1}\right.$ and $\alpha_{2}$ are random! $\alpha_{0}=a$ ) (need 3 points to define a degree 2 polynomial: Âlice, Bob, and Carla each have one point

## Secure Multiparty Computation (yay)

Alice's secret is $A(0)$ where $A()=\alpha_{0} x^{0}+\alpha_{1} x^{1}+\alpha_{2} x^{2} \quad\left(\alpha_{1}\right.$ and $\alpha_{2}$ are random! $\alpha_{0}=a$ ) (need 3 points to define a degree 2 polynomial: Âlice, Bob, and Carla each have one point
$B()=\beta_{0} x^{0}+\beta_{1} x^{1}+\beta_{2} x^{2}\left(\beta_{1}\right.$ and $\beta_{2}$ are random! $\left.\beta_{0}=b\right)$
$C()=\gamma_{0} x^{0}+\gamma_{1} x^{1}+\gamma_{2} x^{2}\left(\gamma_{1}\right.$ and $\gamma_{2}$ are random! $\left.\gamma_{0}=b\right)$

## Secure Multiparty Computation (yay)

Alice's secret is $A(0)$ where $A()=\alpha_{0} x^{0}+\alpha_{1} x^{1}+\alpha_{2} x^{2} \quad\left(\alpha_{1}\right.$ and $\alpha_{2}$ are random! $\alpha_{0}=a$ ) (need 3 points to define a degree 2 polynomial: Alice, Bob, and Carla each have one point
$B()=\beta_{0} x^{0}+\beta_{1} x^{1}+\beta_{2} x^{2}\left(\beta_{1}\right.$ and $\beta_{2}$ are random! $\left.\beta_{0}=b\right)$
$C()=\gamma_{0} x^{0}+\gamma_{1} x^{1}+\gamma_{2} x^{2}\left(\gamma_{1}\right.$ and $\gamma_{2}$ are random! $\left.\gamma_{0}=b\right)$
$A()+B()+C()=D()$ where $D(0)=A(0)+B(0)+C(0)(!!)$
When Alice, Bob, and Carla add their shares locally they obtain a share (a point) of $D()$ that can later be interpolated to learn $D()$ and evaluate $D(0)$ to learn the sum of their secret inputs!

## Secure Multiparty Computation (yay)



## Secure Multiparty Computation (yay)

## Alice (1)

Recover: $d=a+b+c$
Divide $d$ by 3 to get average salary

Bob (2)
Recover: $d=a+b+c$
Divide $d$ by 3 to get average salary

## Carla (3)

Recover: $d=a+b+c$
Divide $d$ by 3 to get average salary

## MPC security model

Alice, Bob, and Carla learn nothing from participating in the protocol that could not have also been learned from only their own input and the protocol output.

## MPC security model

Alice, Bob, and Carla learn nothing from participating in the protocol that could not have also been learned from only their own input and the protocol output.

This is why the computation for average salary outputted the sum of salaries, not the average!

If each participant knows the average salary and the number of participants, they can easily compute the sum of salaries.

## MPC security model

Alice, Bob, and Carla learn nothing from participating in the protocol that could not have also been learned from only their own input and the protocol output.

This is why the computation for average salary outputted the sum of salaries, not the average!

If each participant knows the average salary and the number of participants, they can easily compute the sum of salaries.

No point in computing a sum in 2pc (you'd learn the other party's input!)

## Multiplication in MPC

Multiplication by a constant c can be done locally the same way as addition: multiply the shares $\alpha_{n}$ by $c$ to get a share of a polynomial $D()$ where $D(0)=C \alpha_{0}$

Multiplication of secrets is harder.

## Multiplication in MPC

Multiplication by a constant can be done locally the same way as addition: multiply the shares $\alpha_{n}$ by $c$ to get a share of a polynomial $D()$ where $D(0)=C \alpha_{0}$

Multiplication of secrets is harder.
Firstly, this will raise the degree of the polynomial!
$A()=\alpha_{0} x^{0}+\alpha_{1} x^{1}+\alpha_{2} x^{2}$
$B()=\beta_{0} x^{0}+\beta_{1} x^{1}+\beta_{2} x^{2}$
$A()$ * $B()$ will have degree $4!$ You don't have enough shares to recover that...

## Multiplication in MPC

Multiplication by a constant c can be done locally the same way as addition: multiply the shares $\alpha_{n}$ by $c$ to get a share of a polynomial $D()$ where $D(0)=c \alpha_{0}$

Multiplication of secrets is harder.
Firstly, this will raise the degree of the polynomial!
$A()=\alpha_{0} x^{0}+\alpha_{1} x^{1}+\alpha_{2} x^{2}$
$B()=\beta_{0} x^{0}+\beta_{1} x^{1}+\beta_{2} x^{2}$
$A()$ * $B()$ will have degree 4 ! You don't have enough shares to recover that...
Also, A() * B() will not necessarily be a random polynomial.
Degree reduction and rerandomization is an interactive process that requires communication between Alice, Bob, and Carla.

