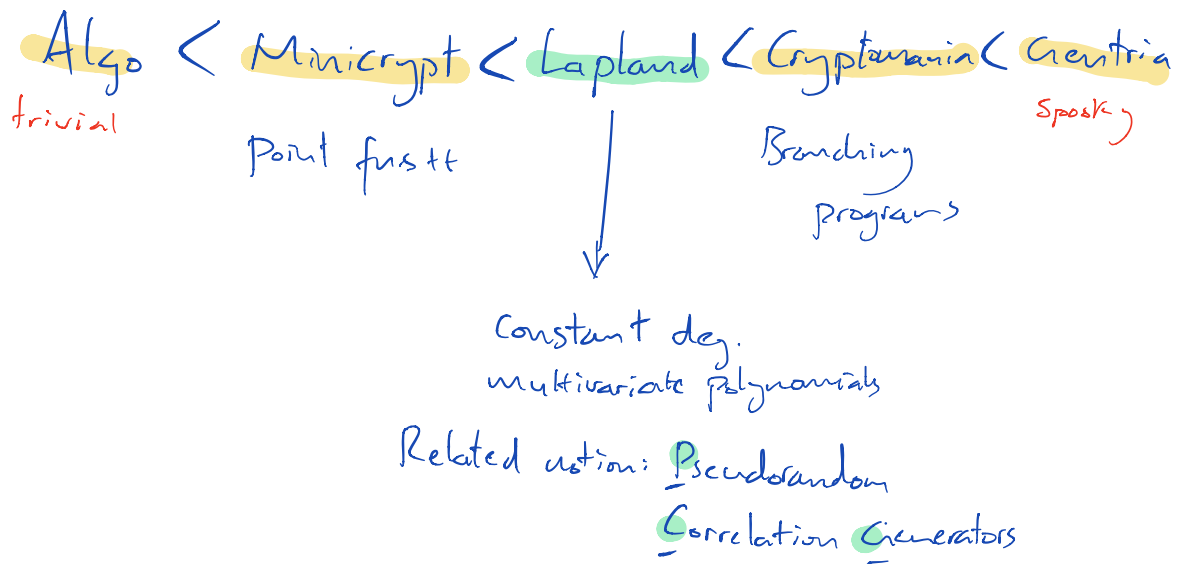


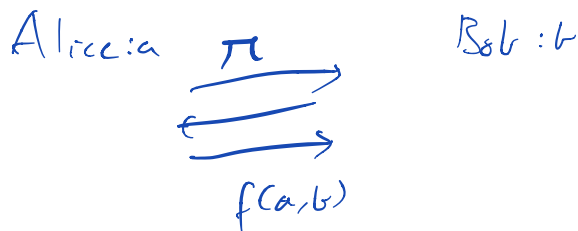
# Worlds of HSS



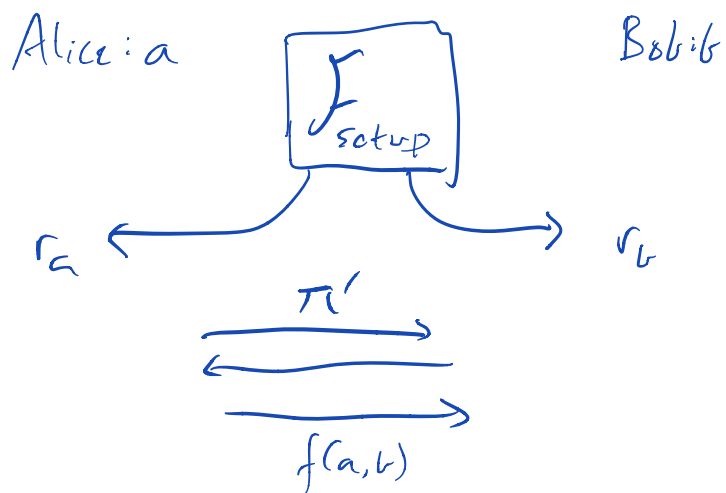
# PCGs

What are "correlations" here?

Recall 2 PC:



## ZPC with correlated randomness



$\pi'$  is cheaper than  $\pi$

ej.  $\downarrow$   
information  
theoretic

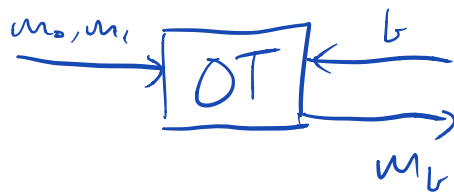
$\downarrow$   
public key crypto

Classic example: Oblivious Transfer  
(OT)

OT : Two parties

S:  $m_0, m_1$

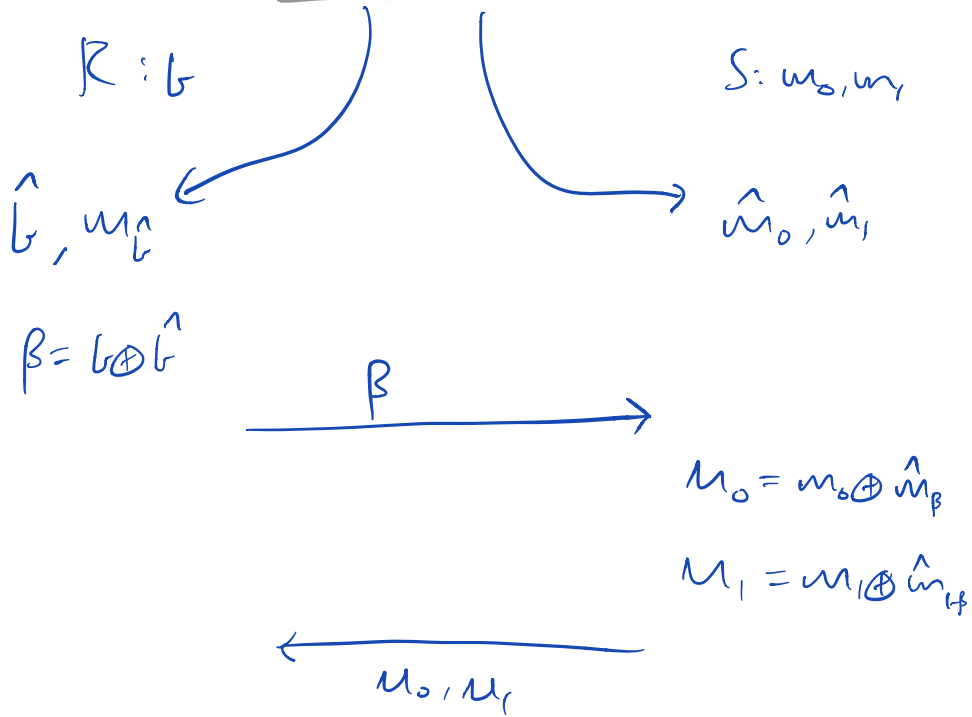
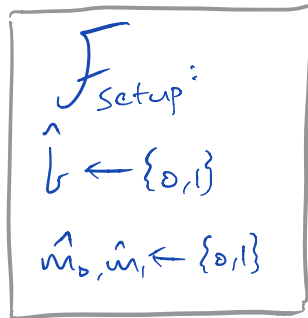
R:  $b \in \{0, 1\}$



OT is complete for secure computation

$\Rightarrow$  cryptographic object, i.e. requires public key operations  
(IR)

But information-theoretic w. preprocessing  
(Beaver 96)



output  $m_b \oplus \hat{m}_b$   
 $= m_b$

Correctness: By inspection

Security:  $\hat{b}$  is OTP for  $b$   
 $\hat{m}_{1-\hat{b}}$  is OTP for  $m_{1-b}$

OT correlation:

$$(\vec{v}, \vec{m}_v), (\vec{w}, \vec{m}_w)$$

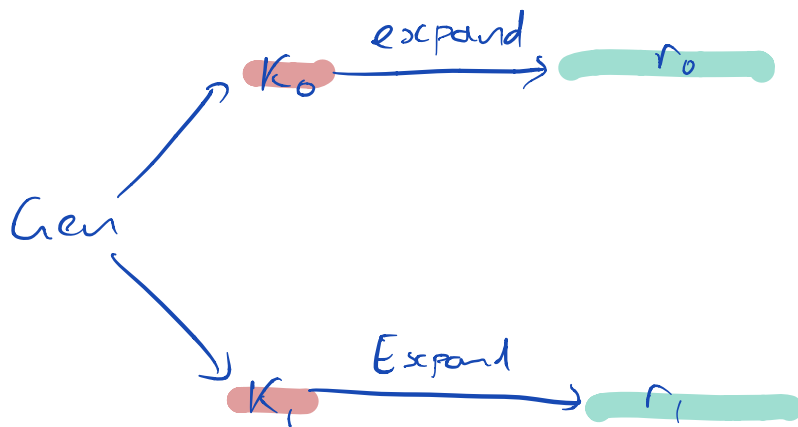
$\Delta$ -OT correlation: [IKNP03]  $\xrightarrow[\text{Hash}]{\text{CorRel}}$  OT

$$\underbrace{(b_i, w_i \oplus b_i \Delta)_{i \in [n]}}_{R_0}, \underbrace{(w_i, w_i \oplus \Delta)_{i \in [n]}}_{R_1}$$

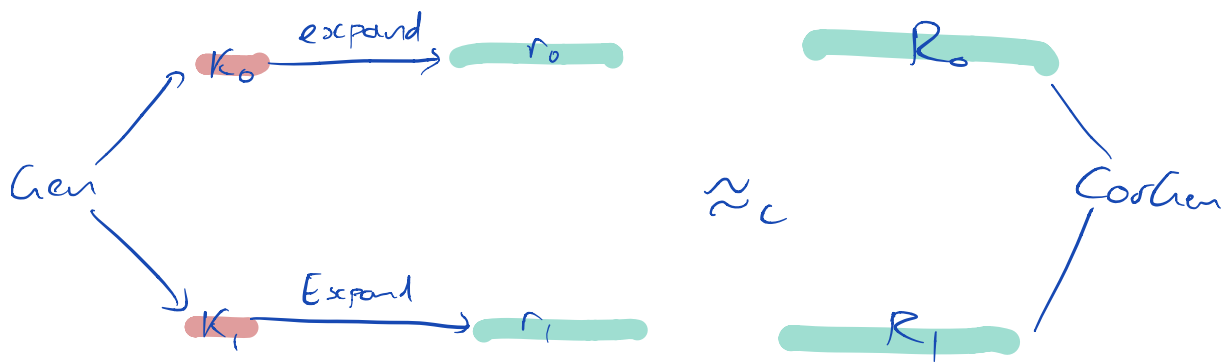
$$R_0, R_1 \leftarrow \text{GenCor}(1^\lambda, n)$$

**Task** compress  $R_0, R_1$  into short seeds  
 $\Rightarrow$  succinct generation of correlations

PCC: Two algorithms



## Correctness:



## Security:

Tricky to define. Simulation-based definition not possible

~~Intuition: due to Yao uncompressible entropy of protocol that simply outputs  $R_0, R_1$  by expanding seeds.~~

Reverse samplability:  $\forall \sigma \in \{0,1\}$

$$R_0, R_1 \leftarrow \text{CorGen}(1^\lambda)$$

$$R'_{1-\sigma} \leftarrow \text{RSample}(\sigma, R_\sigma)$$

$$R'_\sigma = R_\sigma$$

$$(R_0, R_1) \approx_c (R'_0, R'_1)$$

For OT:  $\text{RSample}$ :

Given  $(w_i, w_i \oplus D)$ , sample  $b_i \in \{0,1\}$   
output  $(b_i, w_i \oplus b_i \Delta)$

Given  $(b_i, m_i)$ , sample  $\Delta \in \{0,1\}^\lambda$   
set each  $m_{i-b_i} = m_{i b_i} \oplus \Delta$   
output  $(m_{i0}, m_{i1})$

PCG Security:  $\forall \sigma \in \{0, 1\}$

World 0:

$$k_0, k_1 \leftarrow \text{Gen}(1^\lambda)$$

$$R_\sigma \leftarrow \text{Expand}(\sigma, k_\sigma)$$

output  $(k_{1-\sigma}, R_\sigma)$

World 1:

$$k_0, k_1 \leftarrow \text{Gen}(1^\lambda)$$

$$R_{1-\sigma} \leftarrow \text{Expand}(1-\sigma, k_{1-\sigma})$$

$$R_\sigma \leftarrow \text{Rsample}(G, R_{1-\sigma})$$

output  $(k_{1-\sigma}, R_\sigma)$

Distributions  $\approx_c$

Intuition: captures that other party's  
correlation "as good as" sampled by CorGen  
even when given one key

Turns out to be good enough when setup  
oracle distributes expanded randomness



## Relation to HSS:

Given a PCC for general, additive deg- $d$  correlations (for constant  $d$ ), we can construct an HSS scheme for deg- $d$   $n$ -variate polynomials, whose size is linear in  $n$ .

Not going to cover transformation or PCC for deg- $d$  correlations, but will cover core idea & applications to OT PCC.

Recall OT correlation:

$$R_0 = (\underbrace{w_i}_{\text{Sender}}, \underbrace{w_i \oplus \Delta}_{\text{Receiver}})_{i \in [n]}, \quad R_1 = (\underbrace{b_i}_{\text{Receiver}}, \underbrace{w_i \oplus b_i \Delta}_{\text{Receiver}})_{i \in [n]}$$

Define functions

$$f_s(i) = w_i$$

$$f_r(i) = w_i \oplus b_i \Delta$$

$$(\Delta, f_s) \Rightarrow R_0, \quad (\vec{b}, f_r) \Rightarrow R_1$$

Combining them,

$$f(i) = f_s(i) \oplus f_r(i) = \begin{cases} \Delta & \text{if } b_i = 1 \\ 0 & \text{if } b_i = 0 \end{cases}$$

Look familiar? <sup>Looks like</sup> Multi-point function

But it's not sparse  $\Rightarrow$  unclear how to compress

Dual-LPN : (specific instantiation used here)

Fix integers  $m, n, t$  based on  $\lambda$ .

$HW_{m,t}$  : set of all bit vectors of size  $m$ ,  
and Hamming Weight  $t$

$C$  : probabilistic generation algorithm  
that outputs  $m \times n$  binary matrix  $H$

Assumption:

World 0:

$$H \leftarrow C, e \leftarrow HW_{m,t}$$

$$b = e \cdot H$$

output  $(H, b)$

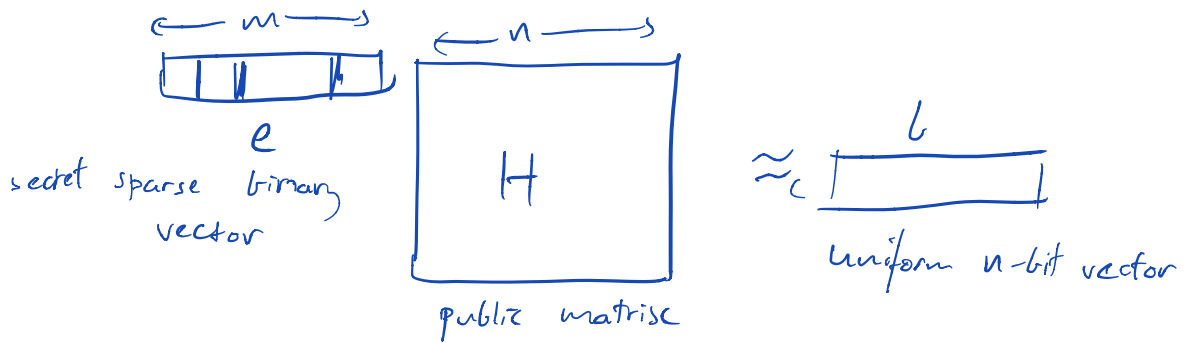
World 1:

$$H \leftarrow C$$

$$b \leftarrow \mathbb{F}_2^n$$

output  $(H, b)$

$$W_0 \stackrel{c}{\approx} W_1$$



$$\text{Each } b_i = \langle e, H_i \rangle$$

↓  
Sparse MPF

→ public weights

↓  
compressible via FSS

→ local evaluation

PCG Gen:

$$\Delta \leftarrow \{0, 1\}^\lambda$$

$$e \leftarrow HW_t$$

$$\text{Set } \hat{f}_{e, \Delta}(j) = \begin{cases} \Delta & \text{if } e_j = 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\hat{f}_0, \hat{f}_1 \leftarrow \text{MPFSS}(\hat{f}_{e, \Delta})$$

$$\text{Output } \underbrace{\kappa_0 = (\hat{f}_0, \Delta)}_{\text{Sender}}, \underbrace{\kappa_1 = (\hat{f}_1, e)}_{\text{Receiver}}$$

Expand  $(\sigma, \kappa_\sigma)$

if  $\sigma = 0$  (Sender):  $(\hat{f}_0, \Delta) := \kappa_0$

Set  $\hat{w}$  s.t.  $(\hat{w}_j = \hat{f}_0(j))_{j \in [m]}$  (full dom. eval)

$$w = \hat{w} \cdot H = (\langle \hat{w}, H_i \rangle)_{i \in [n]}$$

$$R_0 = (w_i, w_i \oplus \Delta)_{i \in [n]}$$

else  $\sigma = 1$  (Receiver):  $(\hat{f}_1, e) := \kappa_1$

$$\text{Set } b = e \cdot H = (\langle e_j, H_i \rangle)_{i \in [n]}$$

$$\hat{v} = (\hat{v}_j = \hat{f}_1(j))_{j \in [m]}$$

$$v = \hat{v} \cdot H = (\langle \hat{v}, H_i \rangle)_{i \in [n]}$$

$$R_1 = (b, v)$$

Correctness:

$$\begin{aligned} v_i &= \langle \hat{v}, H_i \rangle = \sum_{j \in [m]} \hat{f}_1(j) \cdot H_{ij} = \sum (\hat{f}_0(j) \oplus e_j \Delta) \cdot H_{ij} \\ &= \underbrace{(\sum \hat{f}_0(j) H_{ij})}_{w_i} \oplus \Delta \underbrace{\sum e_j H_{ij}}_{b_i} \end{aligned}$$

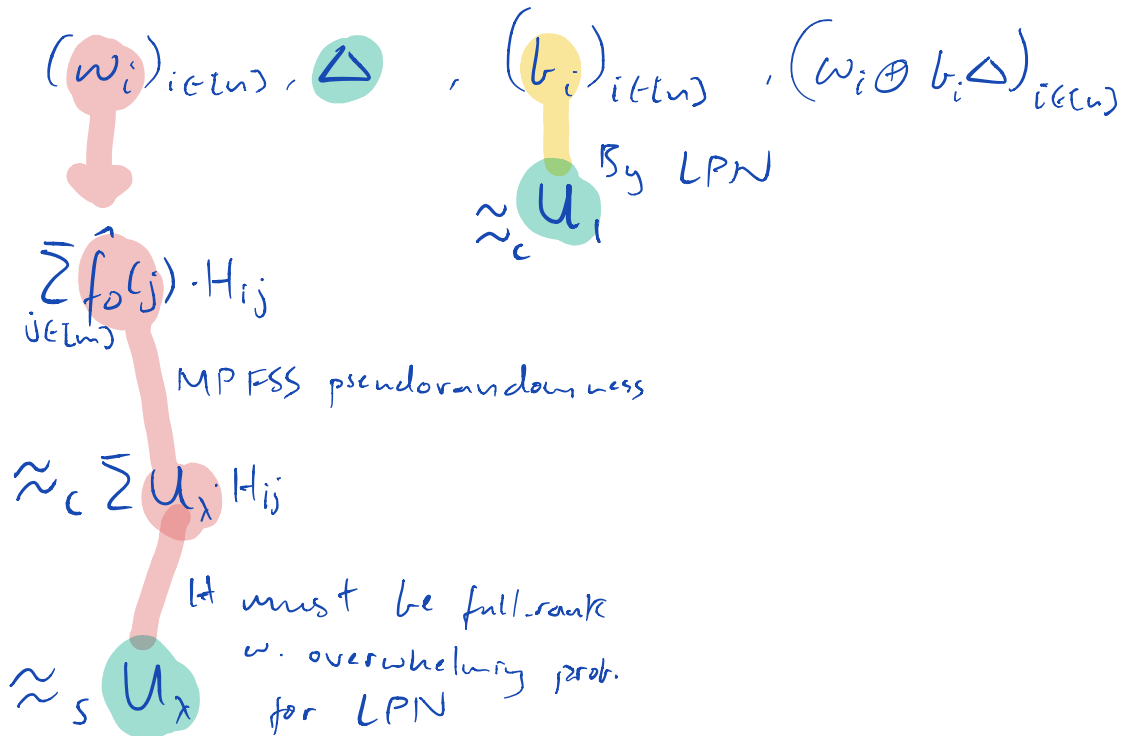
$$v_i = w_i \oplus b_i \Delta$$

Need to argue  $\approx_c (R_0, R_1) \leftarrow \text{Cor Gen}$

$$(w_i)_{i \in [m]}, \Delta, (b_i)_{i \in [m]}, (w_i \oplus b_i \Delta)_{i \in [m]}$$

● uniform

PCA:



# Security:

Receiver: given  $\kappa_r \rightarrow (b_i, w_i \oplus b_i \Delta)$

Ideal:  $(R_{\text{sample}}(R_r))$   
 $\Delta \leftarrow \{0,1\}^\lambda$

$R_o = (w_i, w_i \oplus \Delta)_{i \in [n]}$

Real

$\Delta \leftarrow \{0,1\}^\lambda$

By correctness,  $R_o = (w_i, w_i \oplus \Delta)$

Identical

Conditioned on  $\kappa_r$ ,  $R_o$  is set by  $\Delta$

Sender: given  $\kappa_o \rightarrow (w_i, w_i \oplus \Delta)_{i \in [n]}$

Ideal:  $(R_{\text{sample}}(R_o))$ :

$b \leftarrow \{0,1\}^m$

$R_i = (b_i, w_i \oplus b_i \Delta)_{i \in [n]}$

Real

$e \leftarrow HW_{k,m}$ ,  $b = e \cdot H$

$\hat{f}_0, \hat{f}_1 \leftarrow \text{MPFS}(f_{es})$

$R_i = (b_i, w_i \oplus b_i \Delta)_{i \in [n]}$

depends on  $e$

$\Rightarrow$  Hybrid:  $\hat{f}_0 \leftarrow \text{SimFSS}(1^\lambda)$

Efficiency: (seed size)

Pay to transmit FSS keys for  $t$ -pt. fu.

$$\Rightarrow t \cdot \lambda \log m \in O(\lambda^2 \log \lambda)$$

$\approx \lambda$                        $\lambda^c$

$$\text{Correlation size} = \lambda \cdot m = O(\lambda^{c'+1})$$

$\lambda^c$

Protocol for setup [DS17]:  $O(t \log m)$  OTS

$$\approx O(\lambda^2 \log \lambda)$$

(comm.)