

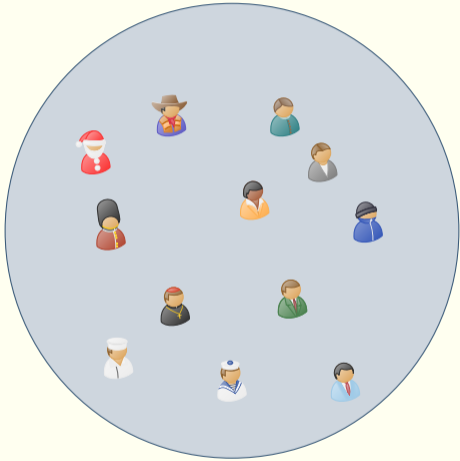
# Jackpot: Non-Interactive Aggregatable Lotteries

Nils Fleischhacker, Mathias Hall-Andersen, Mark Simkin, and Benedikt Wagner

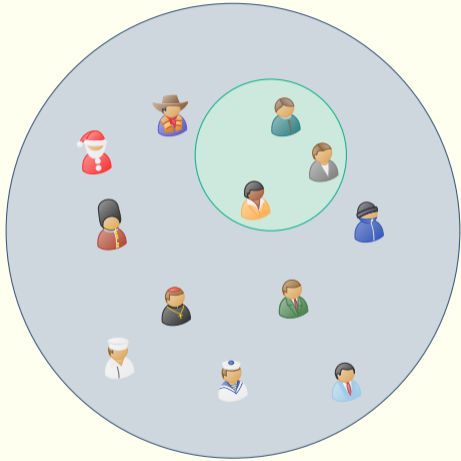
13. December 2024



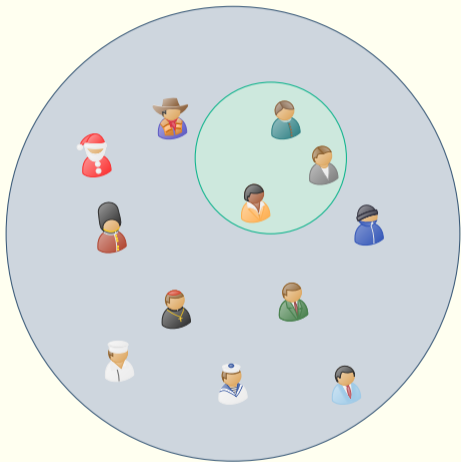
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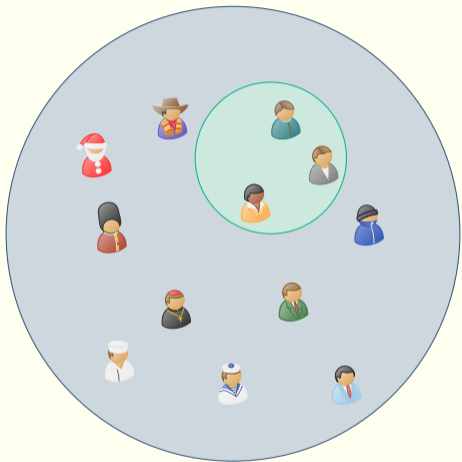
## Why Lotteries?



Goal: Select a committee of size  $\approx N/k$  such that:

- ▶ You can't be selected with probability  $> 1/k$ .

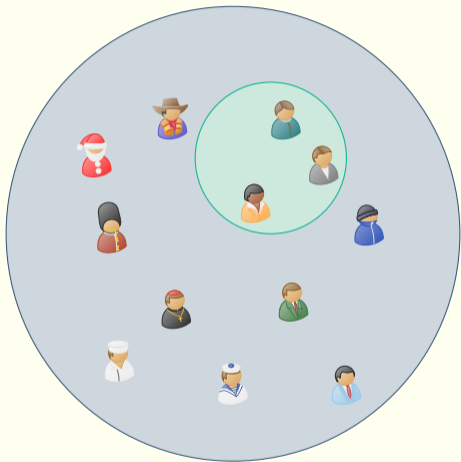
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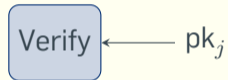
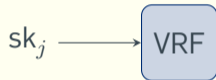
# Why Lotteries?



Goal: Select a committee of size  $\approx N/k$  such that:

- ▶ You can't be selected with probability  $> 1/k$ .
- ▶ Honest participants are selected with  $1/k$ .
- ▶ Participants can't correlate their selection.

## A Lottery Using VRFs

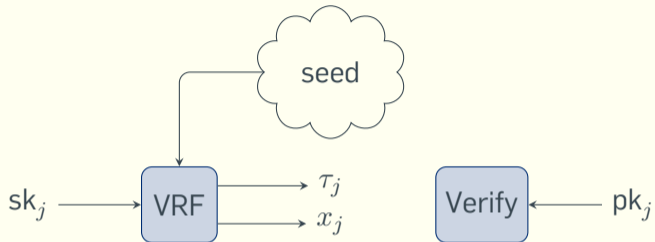


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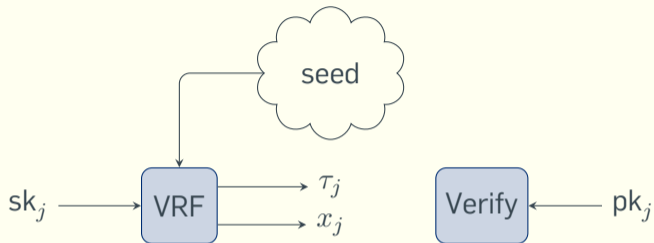




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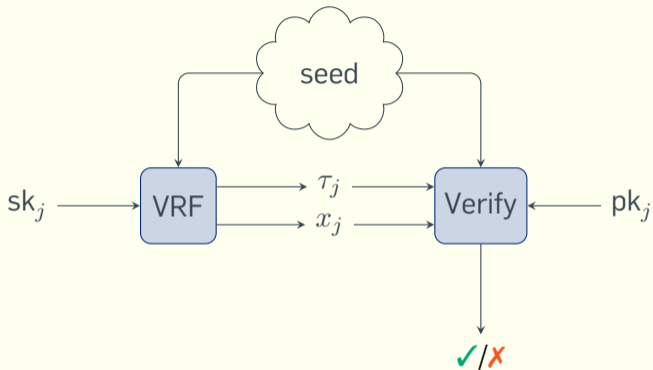


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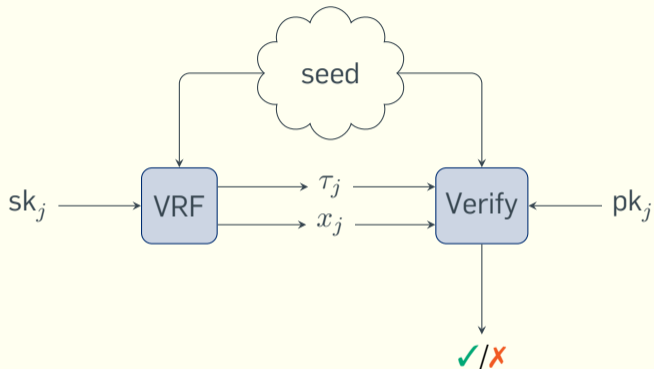
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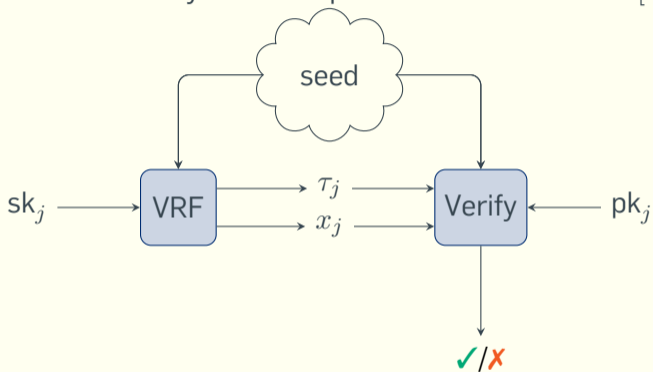


If  $x_j \leq t$  you win the lottery.

Problem: Storing the "winning tickets" takes a lot of space.

## A Lottery Using Small Codomain VRFs

Idea: Use linearly homomorphic VRF with codomain  $[k]$ .



If  $x_j = H(i, j, \text{seed})$  you win the lottery.

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$$\xi := H(i, \text{seed})$$

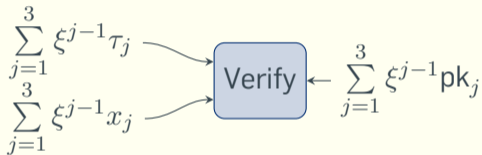




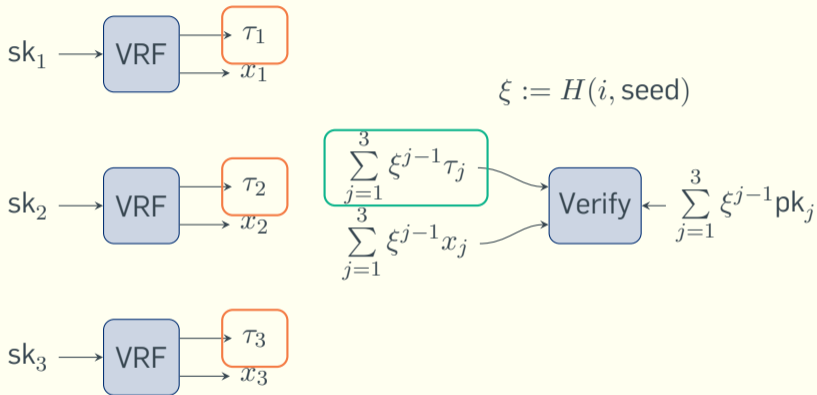
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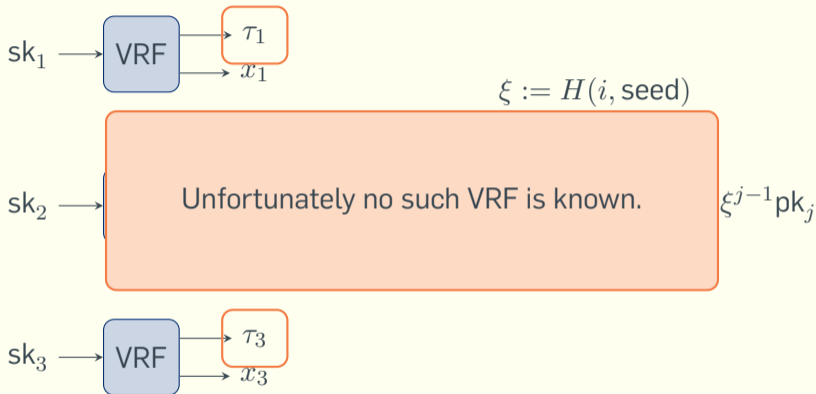
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But what's a VRF other than a vector commitment for long random vectors?

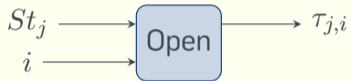


## Using a Vector Commitment

$$(\text{com}_j, St_j) \leftarrow \text{Com}((x_{j,1}, \dots, x_{j,\ell}))$$

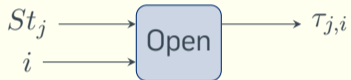
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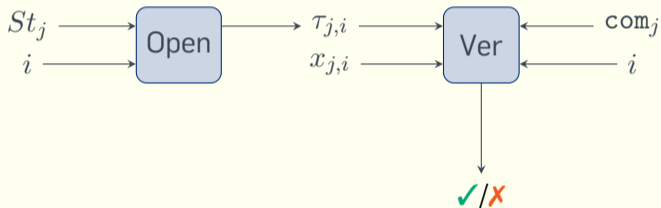
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## Using a Vector Commitment

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Fortunately, we **do** know of linearly homomorphic vector commitments.  
Notably KZG commitments.



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- ▶ But: Only openings have to be aggregateable.
- ▶ Idea: Break homomorphism of commitments but retain it for openings.

## A Simulation Extractable VC from KZG

- ▶ To commit to a vector  $\vec{x}$  of length  $\ell$ :
  1. Choose random polynomial  $f$  of degree  $\ell + 1$  such that  $f(i) = x_i$ .
  2. Compute  $(\text{com}_{\text{KZG}}, St) \leftarrow \text{KZG.Com}(\text{ck}_{\text{KZG}}, f)$ .
  3. Derive  $z_0 = \text{H}(\text{com}_{\text{KZG}})$
  4. Compute  $\tau_0 \leftarrow \text{KZG.Open}(\text{ck}_{\text{KZG}}, St, z_0)$ .
  5. Full commitment is  $(\text{com}_{\text{KZG}}, y_0 = f(z_0), \tau_0)$

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- ▶ We can verify that commitments are well formed
  1. Derive  $z_0 = \text{H}(\text{com}_{\text{KZG}})$ .
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- ▶ Openings are simply openings of the KZG commitment.
- ▶ Openings can be aggregated.
  1. Derive  $\xi := \text{H}'(i, (\text{com}_j)_{j=1}^L, (x_j)_{j=1}^L)$ .
  2. Return  $\tau := \sum_{j=1}^L \xi^{j-1} \tau_j$

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- ▶ Aggregated openings can be verified using the linear homomorphism of KZG.
  1. Individually verify that all commitments are well formed.
  2. Derive  $\xi := \text{H}'(i, (\text{com}_j)_{j=1}^L, (x_j)_{j=1}^L)$
  3. Compute  $x := \sum_{j=1}^L \xi^{j-1} x_j$  and  $\text{com} := \prod_{j=1}^L \text{com}_{\text{KZG}, j}^{\xi^{j-1}}$ .
  4. Return  $\text{KZG.Ver}(\text{ck}_{\text{KZG}}, \text{com}, i, x, \tau)$ .

## Comparison

Tickets $L$	VRF-BLS [B]	Ours [B]	Ratio VRF-BLS/Ours
1	48	80	0.6
16	768	80	9.6
256	12288	80	153.6
1024	49152	80	614.4
2048	98304	80	1228.8

		$L = 1$	$L = 16$	$L = 256$	$L = 1024$	$L = 2048$
Ours	Aggregate [ms]	0.038	0.390	2.377	6.899	14.242
Ours	Ver [ms]	1.413	1.959	3.948	8.875	15.422
VRF-BLS	Ver [ms]	1.663	2.990	7.959	19.010	33.838

