

# An efficient alignment-free method for finding genetic differences between pig races from individual whole genome sequencing data

Jens Zentgraf<sup>†‡</sup> Sven Rahmann<sup>†</sup>

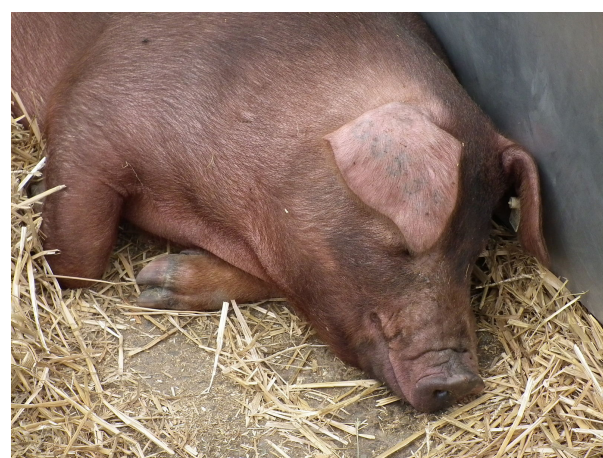
<sup>†</sup> Department of Computer Science and Center for Bioinformatics, Saarland University, Saarbrücken, Germany

<sup>‡</sup> Saarbrücken Graduate School of Computer Science, Saarland University, Saarbrücken, Germany

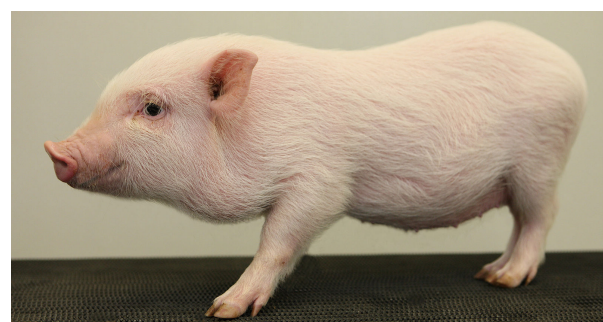
## Goal

Find genes (genomic regions) that are:

- ▶ unaltered between Duroc pigs and Minipigs, lost or heavily altered in Ossabaw
- ▶ explain different physiology in Ossabaw pigs (e.g. obesity)



Duroc



Minipig



Ossabaw

## Limitations and Hardware

- ▶ Only annotated Duroc reference
- ▶ No Minipig or Ossabaw reference with annotations
- ▶ Limited computing resources:
  - ▶ 16 cores (32 threads)
  - ▶ 64 GB RAM

## Datasets

- ▶ **Duroc** reference genome (*Sus Scrofa* reference 11.1)
- ▶ 10 individual sequenced **Minipigs** (NCBI SRA: PRJEB27654, ERR2744277 to ERR2744286)
- ▶ 8 individual sequenced **Ossabaw** pigs (Proprietary data, unpublished, PD Dr. Petra Kleinbongard, Institute of Pathophysiology, University Hospital Essen.)

Species	Samples	Reads (sum)	Gbp (sum)	Cov (sum)
Minipigs	10f	1 969 166 591	393.83	152.65x
Ossabaw	4f	3 349 435 217	1004.83	389.5x
Ossabaw	4m	2 870 736 879	861.22	333.8x

## Classical approach: Alignment

### Disadvantages

- ▶ Requires a lot of computing power
- ▶ Large BAM files

## 1: Workflow for one individual

For ear read in the sample file:

- ▶ Find most plausible origin in the Duroc reference
- ▶ Store alignment in BAM file (large)

## 2: Workflow for one race

For each position in Duroc reference:

- ▶ Find all reads aligned across this position
- ▶ Call race-specific variants

## 3: Compare Minipig and Ossabaw

List all locations with:

- ▶ Variants in Ossabaw race
- ▶ No variant in Minipig race

## References

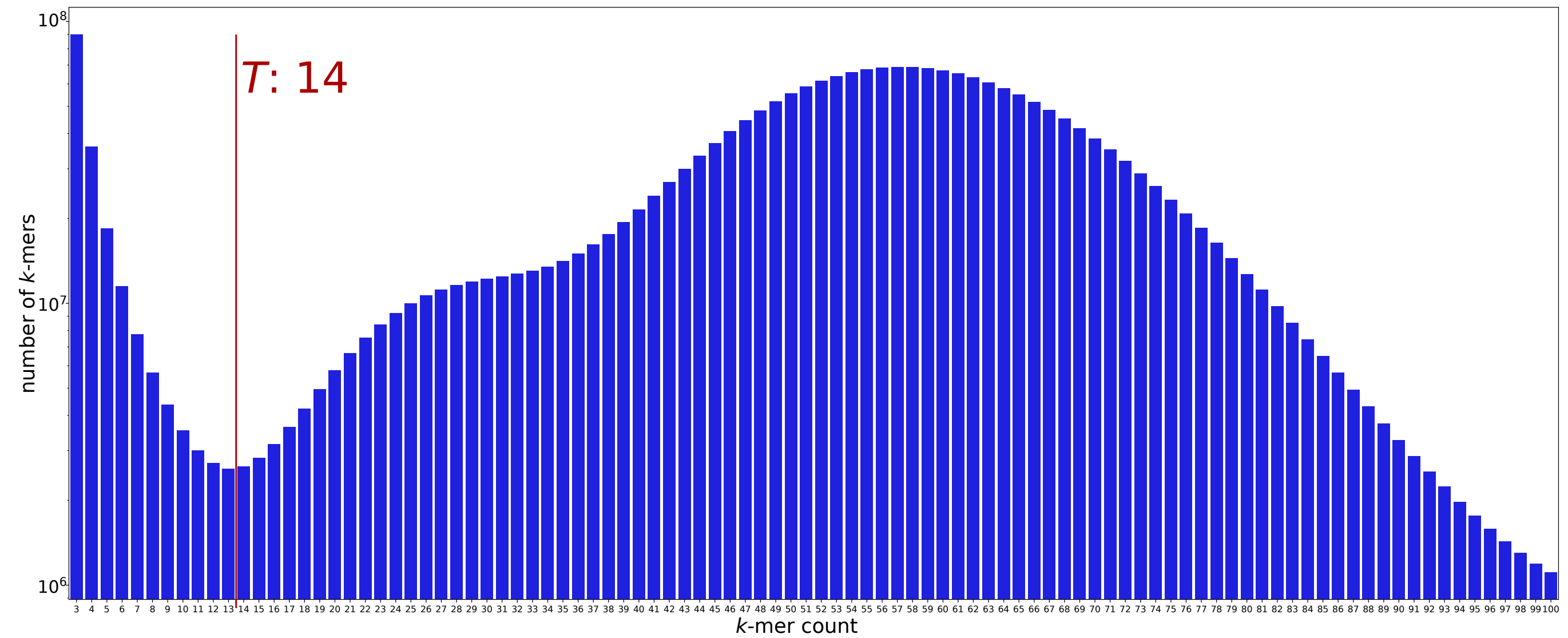
- [1] D. Tang, D. Tan, W. Xiao, J. Lin, and J. Fu. KMC3 and CHTKC: best scenarios, deficiencies, and challenges in high-throughput sequencing data analysis. *Algorithms*, 15(4):107, 2022.
- [2] J. Zentgraf and S. Rahmann. Fast lightweight accurate xenograft sorting. *Algorithms Mol. Biol.*, 16(1):2, 2021.
- [3] J. Zentgraf and S. Rahmann. Fast gapped k-mer counting with subdivided multi-way bucketed cuckoo hash tables. In *22nd International Workshop on Algorithms in Bioinformatics, WABI, 2022*.

## Alignment-free approach

### 1: k-mer Count Histogram (Anarietta, Ossabaw)

Using 3-way bucketed Cuckoo hashing and blocked Bloom filters:

- ▶ Count all  $k$ -mers in the samples
- ▶ Get  $k$ -mer histogram
- ▶ Pick a count Threshold  $T$ :
  - ▶ Count  $\geq T$ :  $k$ -mer belong to the genome
  - ▶ Count  $< T$ :  $k$ -mer is not specific



### 2: k-mer Distribution

$k$ -mer belongs to a species if it belongs to sufficiently many individuals ( $\geq 4$ )

		Minipig										
		0	1	2	3	4	5	6	7	8	9	10
Ossabaw	0	0.00	2.05	1.40	1.26	1.13	1.00	0.96	0.93	0.91	0.94	1.70
	1	0.87	0.08	0.07	0.06	0.06	0.06	0.06	0.06	0.07	0.08	0.15
	2	0.72	0.09	0.08	0.07	0.07	0.07	0.07	0.08	0.09	0.10	0.22
	3	0.80	0.12	0.09	0.10	0.09	0.09	0.09	0.10	0.11	0.13	0.28
	4	1.44	0.21	0.17	0.17	0.17	0.16	0.16	0.17	0.19	0.23	0.52
	5	0.56	0.10	0.09	0.09	0.08	0.08	0.09	0.10	0.11	0.14	0.34
	6	0.61	0.12	0.10	0.09	0.10	0.10	0.11	0.12	0.13	0.15	0.37
	7	0.77	0.18	0.16	0.16	0.16	0.17	0.18	0.20	0.22	0.30	0.85
	8	2.94	1.18	1.15	1.23	1.34	1.48	1.69	2.11	2.91	5.98	49.70

### 3: Comparing Minipig and Ossabaw

For each race:

- ▶ Scan across Duroc reference, one chromosome at a time
- ▶ For each  $k$ -mer look up if:
  - ▶  $k$ -mer is part of the race (**green**)
  - ▶  $k$ -mer is **not** part of the race (**blue**)
- ▶ For each position:
  - ▶ How many green  $k$ -mers cover it (typically  $k$ )

### Variants and lost basepairs

1 2 3 2 1 0 1 2 3 3 3 ...



**Variants:**

- ▶ **Identical regions:** All counters are constant and equal to  $k$
- ▶ **Single nucleotide substitution (1bp):** Counter drop from  $k$  to 0, then go back up to  $k$
- ▶ **Isolated deletion of length  $\ell$ :**
  - ▶ Counter drops from  $k$  to 0
  - ▶ Stay at 0 for  $\ell$  base pairs
  - ▶ Go back up to  $k$

**Lost basepair:**

- ▶ Position with small counter (0 if strict)

### Complications

- ▶  $k$ -mers may appear in different places in the genome, not only at the current position
  - ▶ Variants may not be isolated
- ⇒ Complex, hard to interpret counter pattern
- ▶ We do **not** need individual genome variants

### Preliminary Results

High-ranked genes: mitochondrial proteins, inflammatory proteins from JAK-STAT pathway

## Discovery Method

- ▶ Consider exons of genes of the Duroc reference
- ▶ For each gene: Count number of lost Duroc basepairs in Minipig and Ossabaw genome
- ▶ Interesting regions depends on threshold  $T$ 
  - ▶ High loss in Ossabaw
  - ▶ low (zero) loss in Minipig
- ▶ Analyze gene list

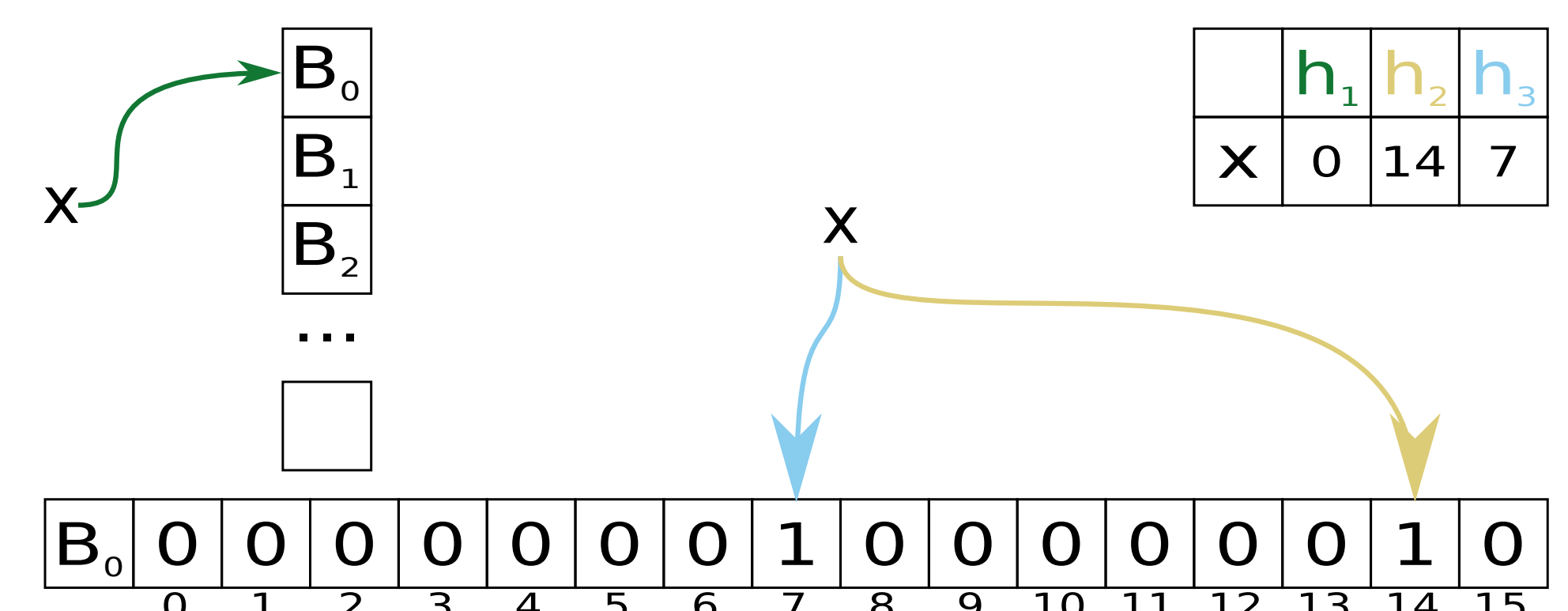
## Blocked Bloom Filter

**Problem:** Very many rare  $k$ -mers

- ▶ In total 40 to 50 G  $k$ -mers occur only once or twice
- ▶ Only interested in the  $\leq 2.5$ G frequent  $k$ -mers
- ▶ Only 64 GB of RAM to store  $k$ -mers and counters

**Solution:** Use multiple Bloom Filters

- ▶ 2 blocked Bloom filters
- ▶ First catches all  $k$ -mers that occur once
- ▶ Second catches all  $k$ -mers that occur twice



## 3-Way Cuckoo Hash Table

Cuckoo Hashing:

- ▶  $h = 3$  hash functions
- ▶  $b = 4$  bucket size
- ▶ A key has  $hb$  possible slots
- ▶ If full, replace random key and reinsert old key
- ▶ Repeat for fixed number of steps, FAIL if not possible

Save memory:

- ▶ High load  $\geq 90\%$  (up to 99.9%)
- ▶ Use word packing (only 54 bits for a 27-mer, not a full uint64)
- ▶ Only store quotient of a  $k$ -mer

Save time:

- ▶ Use multiple threads
- ▶ Write access to the same memory
  - ▶ Use independent sub-tables
  - ▶ All hash functions map to same subtable
  - ▶ Use producer-consumer model

