

# Modern hashing for alignment-free sequence analysis



## Part 1: Introduction (*k*-mers, alignment-free methods)

Jens Zentgraf & Sven Rahmann  
GCB 2021



# Tentative time table

1. Introduction: motivation,  $k$ -mers, alignment-free methods  
(09:00 - 09:45, Sven)
2. Hashing, hash functions, collision resolution  
(09:45 - 10:30, Jens)

Short break (20 min)

3. Multi-way bucketed cuckoo hashing for DNA  $k$ -mers  
(10:50 - 11:30, Jens)
4. Performance engineering  
(11:30 - 12:00, Sven)

# Foundation of most DNA sequence analysis tasks in bioinformatics

**1. Read mapping:** Find genomic origin(s)  
of a given DNA sequence (the "read")

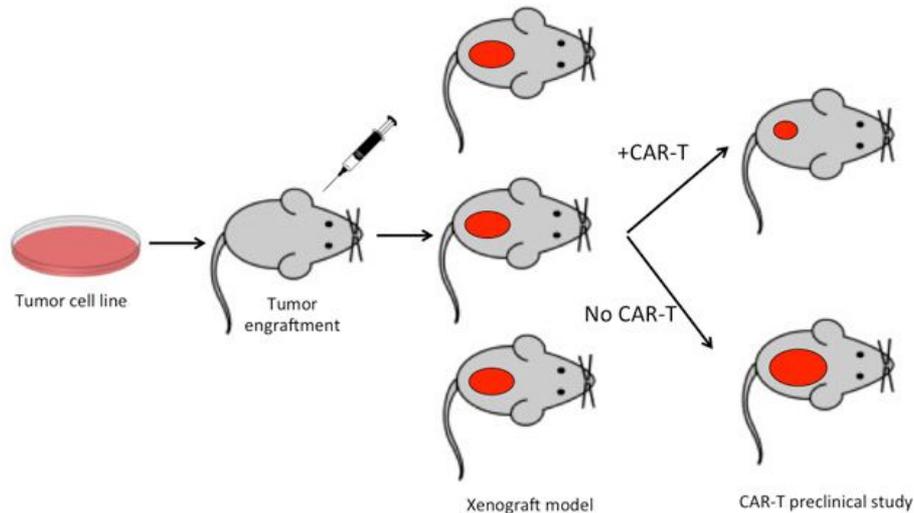
**2. Read alignment:** Base-by-base comparison of read and genome  
(often mingled together, but really 2 distinct steps!)

**This tutorial:** How to short-cut mapping and avoid alignment

- Find all exact occurrences of short  $k$ -mers (DNA substrings of length  $k$ )
- Do this fast, for billions for  $k$ -mers

Motivation: Xenograft sorting

# (Patient-derived) xenografts



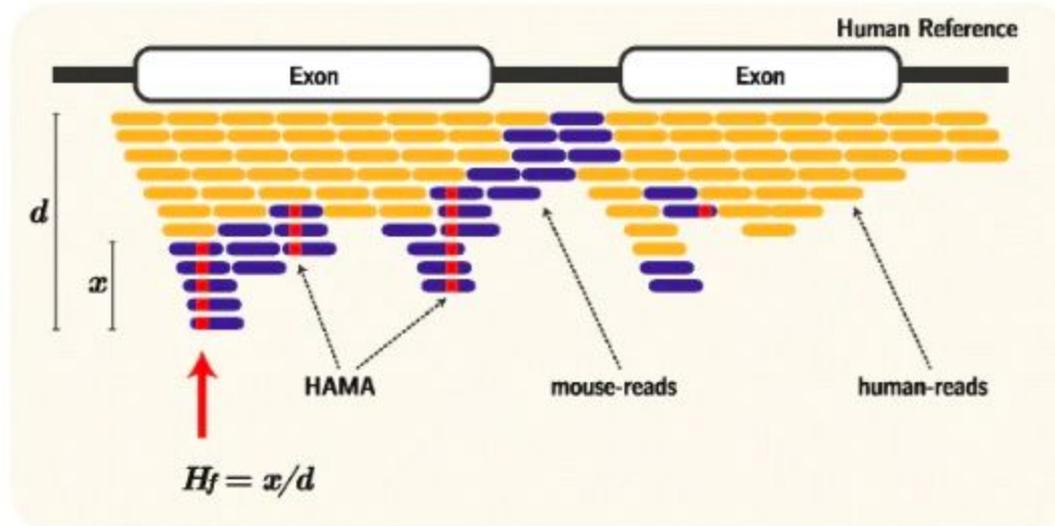
- tumor cell lines or patient tumor samples implanted in mice
- study tumor heterogeneity, evolution
- sequencing of samples
- mixture of human+mouse DNA
- First task: separate/sort reads ("xenograft sorting"), or: extract graft (human) reads

Source: Creative AniModel,

<https://www.creative-animodel.com/Featured-Service/Human-Tumor-Xenograft-Model.html>

# Problem: Human-Aligned Mouse Alleles (HAMAs)

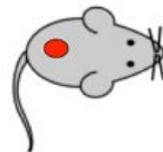
- mouse reads may align to human genome
- may lead to false human (tumor) variant calls
- oncogenes particularly prone to this effect



S. Y. Jo, E. Kim, and S. Kim.  
**Impact of mouse contamination in genomic profiling of patient-derived models and best practice for robust analysis.**  
*Genome Biology*, 20(1):Article 231, Nov 2019.

# The xenograft sorting problem

**Given:** sequenced xenograft sample (reads from two species),  
paired-end or single-end,  
genomic or transcriptomic reads,



**sort** the reads into five categories according to species of origin:

host (mouse), graft (human), both, neither, ambiguous

or: **partially sort** using fewer categories (host, graft, other),

or: **count** how many reads are in each category,

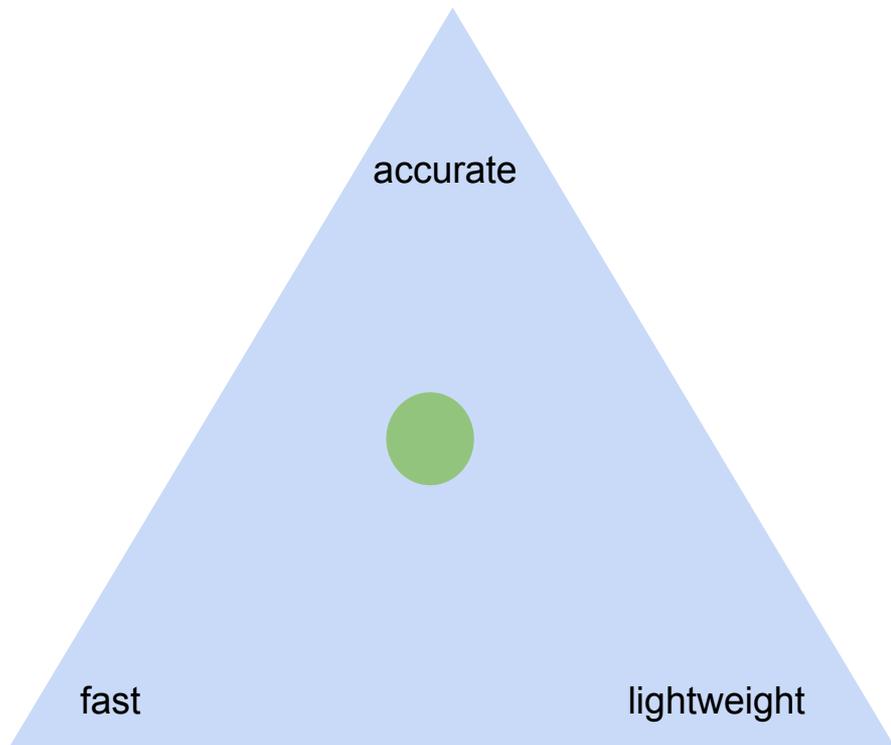
or: **filter** (select) only graft (human) reads.

# *k*-mer methods for xenograft sorting

- Partition each **read** into its *k*-mers
- Look up information on each *k*-mer in hash table  
[*k*-mer ↦ **human** | **mouse** | **both**]
- Absent *k*-mers occur in **neither** species.
- **Aggregate** *k*-mer information into a statement about the **read** (majority vote, complex decision rule, ...).

GATTCATGC . . .  
GATTC  
ATTCA  
TTCAT  
TCATG  
CATGC  
 . . . . .

# Goal: "Fast lightweight accurate xenograft sorting"



## fast:

- slow random memory accesses
- 3-way bucketed Cuckoo hashing
- buckets fit within a cache line

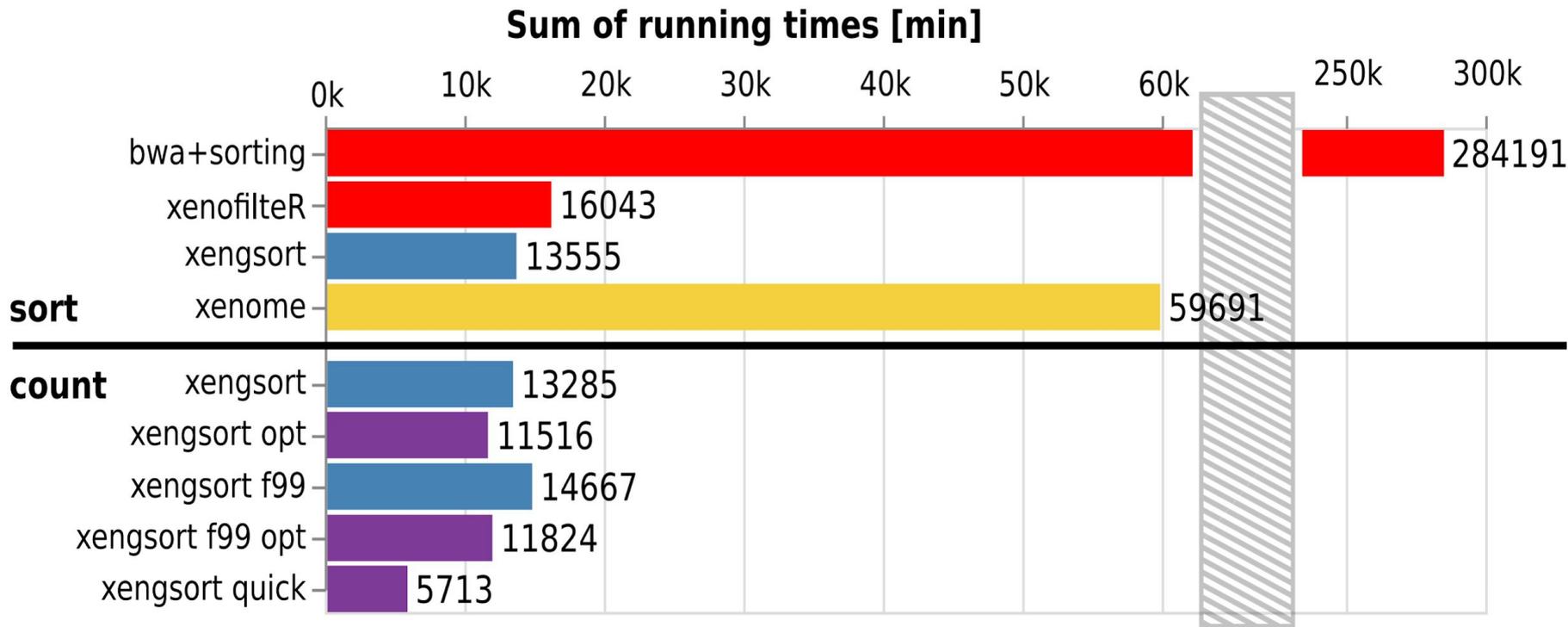
## lightweight (small memory footprint):

- 4.5 billion 25-mers + values
- high load (little wasted space)
- quotienting

## accurate:

- identical + highly similar sequences
- "weak"  $k$ -mers
- multi-level decision rule

# 174 PDX datasets: Running times [CPU minutes]



# Examples of alignment-free methods

1. **Xengsort:** Xenograft sorting (already discussed)  
(<https://gitlab.com/genomeinformatics/xengsort> - 2020)
2. **BCOOL:** Sequencing error correction  
(<https://github.com/Malfoy/BCOOL> - 2019)
3. **Kraken 2:** metagenomic species identification and quantification  
(<https://ccb.jhu.edu/software/kraken2/> - Sep 2019)
4. **Kallisto:** RNA-seq transcript quantification  
(<https://pachterlab.github.io/kallisto/> - 2016);  
not for differential expression; use additional tools like sleuth
5. **DE-kupl:** discovery of novel (differentially expressed) transcripts  
(<https://transipedia.github.io/dekupl/> - 2017)

# K-mers and their encodings, Minimizers and sketches

# $k$ -mers and their integer encodings

**$k$ -mer:** any DNA/RNA sequence of length  $k$ .

There are  $4^k$  different DNA  $k$ -mers.

**Other names:**  $k$ -mer,  $q$ -gram,  $n$ -gram,  $\ell$ -mer, shingle, ...

**$k$ -mer code / encoded  $k$ -mer:** Translating **A=0**, **C=1**, **G=2**, **T=3**

(or any other bijective map  $\{A,C,G,T\} \rightarrow \{0,1,2,3\}$ ) for fixed  $k$ ,

a  $k$ -mer becomes an integer (base-4 number) in  $\{0, 1, \dots, 4^k-1\}$ .

**Example:** **TATCG**  $\mapsto$  **(30312)**<sub>4</sub> = **3** · 256 + **0** · 64 + **3** · 16 + **1** · 4 + **2** · 1 = 822

# Canonical $k$ -mers

**canonical  $k$ -mer:** DNA is **double-stranded**;

a  $k$ -mer is the same molecule as its reverse complement,  
the canonical representation is the **lexicographically smaller** one.

**Example:** TATCG = CGATA, canonical: CGATA.

**canonical code:** integer code of canonical  $k$ -mer

minimum of encodings of  $k$ -mer and its reverse complement;  
always need to evaluate both  $k$ -mer  $x$  and  $rc(x)$ .

**Example:**  $\text{code}(\text{TATCG}) = \text{code}(\text{CGATA}) = \min(822, 716) = 716$ .

**Note:** works equally well with  $\text{max}()$  instead of  $\text{min}()$

# Contiguous vs. gapped $k$ -mers

## contiguous $k$ -mer (standard):

$k$ -mer that occurs as one contiguous substring

## gapped / spaced $k$ -mer and mask:

- gap pattern given by (symmetric!) mask: e.g.: #\_\_#\_\_#\_\_#
- #: significant positions ( $k$ ) vs. \_: gap positions / spacers ( $s$ )
- $k$ -mer by concatenating significant positions (**weight**  $k$ , **span/width**  $w = k+s$ )
- advantage: cover sequence width in fewer steps

**Example:** AGGTCGGTAGGC

#\_#\_#\_#\_# ATGG

#\_#\_#\_#\_# GCTG

#\_#\_#\_#\_# GGAC

AGGTCGGTAGGC

#####

#####

#####

3  $k$ -mers cover

12 positions (gapped)

6 positions (cont.)

# Key-value stores

## General definition:

A **key-value store** ("key-value database") is a data structure that stores objects or records ("values"), each of which is associated to an immutable "key" object.

## Examples:

- Java HashMap
- Python dict
- Databases: redis, Oracle NoSQL, memcached, ...

## Restricting values:

Values in key-value databases may be any object, even with different types!  
Keys can be any immutable hashable object (often strings or tuples of numbers).

**We assume** that the value type is known and fixed (value set  $V = \{0, \dots, |V|-1\}$ , so values have fixed bit width (e.g. 32 bits).

(Circumvented by storing pointers to arbitrary objects -- what the databases do anyway)

# Minimizers

**Given:** Two integers  $k \leq w$  ( $w$ : "window width" in a DNA sequence)

**Definition:** A (canonical)  $k$ -mer  $m$  is a **minimizer** in a window of length  $w$  iff

- $m$  is a (canonical)  $k$ -mer in the window,
- its (canonical) code is the **smallest** of all (canonical) codes in the window.
- The "smallest" may be with respect to a permutation of  $k$ -mers.

## Advantages of minimizers

- Minimizers tend to stay (locally) constant for overlapping windows
- There are fewer different minimizers than windows
- Similar sequences have high probability of having the same minimizer(s).
- Sequence of minimizers is also called a **sketch** of the original sequence.

# Minimizers

Example:  $k = 3$ ,  $w = 6$  (4  $k$ -mers), AGGTCGGTAGGC

k-mer	cc <sub>max</sub>	minim(6)
AGG	23 (CCT)	23
GGT	43	26
GTC	45	26
TCG	54	26
CGG	26	26
GGT	43	23
GTA	49	23
TAG	50	-/-
AGG	23	-/-
GGC	41	-/-

cc <sub>max</sub>	cc xor (101010)	minim(6)
23 (CCT)	61	1
43	1	1
45	7	1
54	28	1
26	48	1
43	1	1
49	27	3
50	24	-/-
23	61	-/-
41	3	-/-

## Note:

xor-ing canonical codes with random numbers and taking the minimum "simulates" different random permutations of numbers w.r.t. taking the minimum.

$$\begin{aligned} 43 &= (101011)_2 \\ \text{xor } (101010)_2 \\ &= 1 = (000001)_2 \end{aligned}$$

# Data structures for key-value-stores (in memory)

Two basic possibilities to look up keys fast:

- **sorting** (binary search)
  - variants of lists (e.g. skip lists)
  - (balanced) search trees
- **hashing** (compute an address / index in an array)
  - typically arrays, but may need to be re-sized
  - collisions must be resolved
- **hybrids** (binning/hashing by prefix, sorted within bin)

**Note:** on small datasets, do nothing, linear scan is fast enough!

# Applications of the alignment-free paradigm

# Examples of alignment-free methods

1. **Xengsort:** Xenograft sorting (already discussed)  
(<https://gitlab.com/genomeinformatics/xengsort> - 2020)
2. **BCOOL:** Sequencing error correction  
(<https://github.com/Malfoy/BCOOL> - 2019)
3. **Kraken 2:** metagenomic species identification and quantification  
(<https://ccb.jhu.edu/software/kraken2/> - Sep 2019)
4. **Kallisto:** RNA-seq transcript quantification  
(<https://pachterlab.github.io/kallisto/> - 2016);  
not for differential expression; use additional tools like sleuth
5. **DE-kupl:** discovery of novel (differentially expressed) transcripts  
(<https://transipedia.github.io/dekupl/> - 2017)

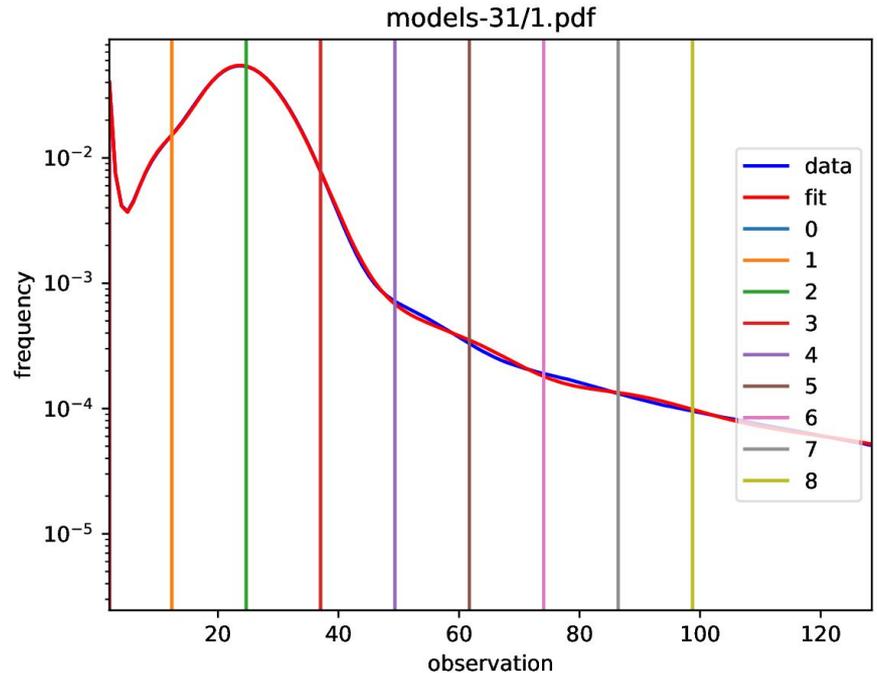
# BCOOL - sequencing error correction

Software: <https://github.com/Malfoy/BCOOL>

Papers: arXiv: <https://arxiv.org/pdf/1711.03336.pdf>;

## Ideas:

- count number of occurrences of each k-mer in all reads
- build k-mer histogram
- k-mers occurring rarely are probably errors and must be corrected
- Reads are mapped to De Bruijn graph



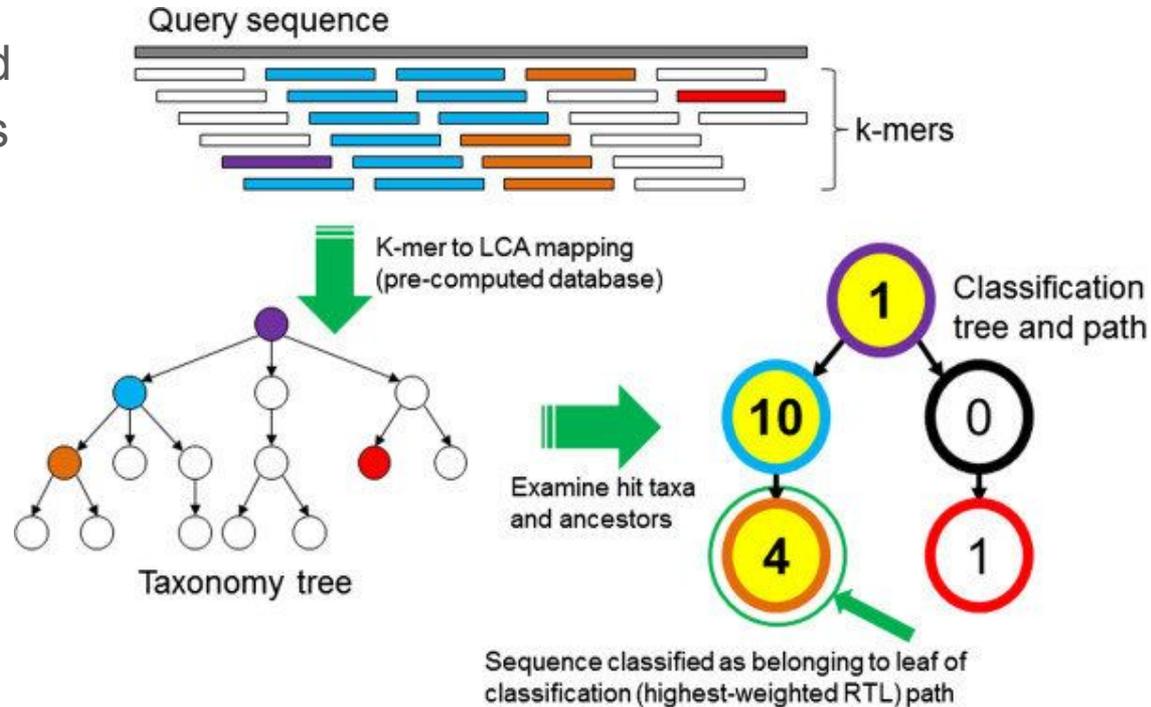
# Kraken 2 - Metagenomic species identification

**Software:** <https://ccb.jhu.edu/software/kraken2/>

**Preprint:** <https://www.biorxiv.org/content/10.1101/762302v1>

Examines  $k$ -mers of a query read to find the most probable species of origin in a taxonomy tree.

Each  $k$ -mer is mapped to a tree node (lowest common ancestor, LCA) of all species containing the  $k$ -mer.



# Kallisto: RNA-seq transcript quantification

Software: <https://pachterlab.github.io/kallisto/>

Paper: <https://www.nature.com/articles/nbt.3519>

## Ideas:

- map each k-mer of a read to a set ("compatibility class") of transcripts (typically from .cdna.fasta files)
- take (soft) intersection of compatibility classes (perhaps do read error correction before mapping)
- run a decoding algorithm on reads that cannot be uniquely placed

# DE-kupl: Discovery of new differential transcripts

**Software:** <https://transipedia.github.io/dekupl/>

**Paper:** <https://genomebiology.biomedcentral.com/articles/10.1186/s13059-017-1372-2>

## Ideas:

- Count occurrence of each k-mer in RNA-seq datasets (from two classes)
- Remove k-mers from known transcripts
- Do test of differential expression on remaining ("novel") k-mers
- Locally assemble differential novel k-mers;  
yields novel differentially expressed transcripts (parts of them)

Next part:  
Hashing, hash functions  
collision resolution