## The Fine-Grained Complexity of Episode Matching

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## Episode Matching

$P=$ ANANAS
$S=\underset{0}{\text { BATMAN }} \operatorname{AND}_{10}$ ANNA $_{15} \operatorname{SING} \underset{20}{\text { NANANANA }} \underset{25}{\text { AND }} \underset{35}{\text { EAT }} \underset{40}{\text { BANANAS }}$

## Episode Matching

$$
\begin{aligned}
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- Find minimal substrings of $S$ containing $P$ as a subsequence


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- The minimal substrings of $S$ which contain $P$ as a subsequence are shown in blue: $S[6,16]$ and $S[39,44]$


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- Find minimal substrings of $S$ containing $P$ as a subsequence
- The minimal substrings of $S$ which contain $P$ as a subsequence are shown in blue: $S[6,16]$ and $S[39,44]$
- We consider a version of the problem where the goal is to find the length of the shortest substring of $S$ containing $P$ as a subsequence


## Complexities - Algorithms

- $|P|=m,|S|=n$


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- This work: no $O\left(n m^{1-\epsilon}\right)$ or $O\left(n^{1-\epsilon} m\right)$ algorithm assuming OVH


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| [AA02] | $O(n)$ | $O\left(\sum_{i=1}^{m} \operatorname{dist}\right.$ _occ $\left.\left(P_{i}\right) \cdot i\right)$ |  |
| This work | $O\left(n+\left(\frac{n}{\tau}\right)^{k}\right)$ | $O(k \cdot \tau \cdot \log \log n)$ | $m=k$ fixed |
| This work | $\Omega\left(n^{k-k \delta-o(1)}\right)$ | $O\left(n^{\delta}\right)$ | $m=k$ fixed |

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- dist_occ $\left(P_{i}\right)$ is the number of distinct minimal substrings containing $P[1] \ldots P[i]$ as a subsequence
- Conditional lower bound based on hardness of $k$-Set Disjointness


## Complexities - Special case $|P|=2$

- This work: Faster preprocessing for decision version using min-plus matrix multiplication


## Orthogonal Vectors

- Two sets $A, B$ of $d$-dimensional, binary vectors, each set has size $n$
- Problem: Decide if there is a vector in $A$ that is orthogonal to
a vector in $B$
- OVH: There is no algorithm running in time $O\left(n^{2-\epsilon} \operatorname{poly}(d)\right)$


## OV $\rightarrow$ Episode Matching

- build $P$ from $B$ : for $b \in B$, seperate each coordinate by new letter $x$
eg: $101 \rightarrow 1 \times 0 \times 1$


## UV $\rightarrow$ Episode Matching

- build $P$ from $B$ : for $b \in B$, separate each coordinate by new letter $x$
eg: $101 \rightarrow 1 \times 0 \times 1$
- concatenate and separate by new letter $y$
eg: $B=\{101,111,110\}$,
$P=1 \times 0 x 1 y 1 x 1 x 1 y 1 x 1 x 0$


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$P=1 \times 0 \times 1 y 1 \times 1 \times 1 y 1 x 1 x 0$
- Length of $P=O(n d)$


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- build $S$ from $A$ : for $a \in A$,

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& 1 \rightarrow 00
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separate each coordinate by letter $x$
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- $0 \times 1 \times 0$ is a subsequence of $00 \times 01 \times 01$
- $b_{2}$ and $a$ are not orthogonal
- $1 \times 1 \times 0$ is not a subsequence of $00 \times 01 \times 01$


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s\left(a_{1}\right) y s(z) y s\left(a_{2}\right) y s(z) y \ldots s\left(a_{n}\right) y s(z) y s\left(a_{1}\right) y s(z) y \ldots s(n)
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- Length of $S=O(n d)$
- $|P||S|^{1-\epsilon}=O\left(n^{2-\epsilon} d^{2-\epsilon}\right)$
$|P|^{1-\epsilon}|S|=O\left(n^{2-\epsilon} d^{2-\epsilon}\right)$


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No orthogonal vectors:

$$
\begin{array}{llllllllll}
y & s(z) & y & s\left(a_{i-1}\right) & y & s(z) & y & s\left(a_{i}\right) & y & s(z) \\
y & p\left(b_{j-1}\right) & & y & p\left(b_{j}\right) & & & y\left(a_{i+1}\right) \\
& & p\left(b_{j+1}\right) &
\end{array}
$$

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$a_{i}, b_{j}$ orthogonal:
$\begin{array}{llllllllll}y & s(z) & y & s\left(a_{i-1}\right) & y & s(z) & y & s\left(a_{i}\right) & \text { y } & s(z) \\ y & p\left(b_{j-2}\right) & & & \text { y } p\left(b_{j-1}\right) & \text { y } s\left(a_{i+1}\right) \\ p\left(b_{j}\right) & \text { y } p\left(b_{j+1}\right) & \end{array}$

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$$
S=s\left(a_{1}\right) y s(z) y s\left(a_{2}\right) y s(z) y \ldots s\left(a_{n}\right) y s(z) y s\left(a_{1}\right) y s(z) y \ldots s\left(a_{n}\right)
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- $a_{i} \perp b_{j}$
- $j<i$ : "overflow" to the right
- $j>i$ : "overflow" to the left


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## Binary alphabet

- replace $x$ and $y$ by binary gadgets


## Space/time trade-off

- $|P|=k$ fixed at preprocessing
- Upper bound: Space: $O\left(n+\left(\frac{n}{\tau}\right)^{k}\right)$, Time: $O(k \cdot \tau \cdot \log \log n)$ $m=k$
- Conditional lower bound: Space: $\Omega\left(n^{k-k \delta-o(1)}\right)$, Time: $O\left(n^{\delta}\right)$


## Space/time trade-off, Lower bound

Definition ( $k$-Set Disjointness Problem)
Preprocess $m$ sets $S_{1}, S_{2}, \ldots, S_{m}$ of total size $\sum_{i=1}^{m}\left|S_{i}\right|=N$ drawn from a universe $U$ such that given ( $i_{1}, i_{2}, \ldots, i_{k}$ ) we can quickly decide whether $\bigcap_{j=1}^{k} S_{i j}=\emptyset$.

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Conjecture (Strong $k$-Set Disjointness Conjecture)
Any data structure for the $k$-Set Disjointness Problem that answers queries in time $T$ must use $\tilde{\Omega}\left(N^{k} / T^{k}\right)$ space.

## Space/time trade-off, Lower bound

$$
\begin{array}{ll}
S_{1}=\{1,3,4\} & \alpha_{1} \\
S_{2}=\{2\} & \alpha_{2} \\
S_{3}=\{1,2,3,4\} & \alpha_{3} \\
S_{4}=\{2,4\} & \alpha_{4} \\
S_{5}=\{1,3\} & \alpha_{5}
\end{array}
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$S_{1} \cap S_{4}=\emptyset ?$

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$S_{2} \cap S_{5}=\emptyset ?$

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$$
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$$

$$
S_{2} \cap S_{5}=\emptyset ?
$$

$$
P_{2}=\alpha_{2} \alpha_{5}
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- Space $=O\left(n+\left(\frac{n}{\tau}\right)^{k}\right)$, Time $=O(k \cdot \tau \cdot \log \log n)$


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- Space $=O\left(n+\left(\frac{n}{\tau}\right)^{k}\right)$, Time $=O(k \cdot \tau \cdot \log \log n)$
- Call letters appearing more than $\tau$ times frequent
- For all $k$-tuples of frequent letters precompute answers
- Have a predecessor data structure for each letter (total size $=$ $O(n))$
- If $P$ contains non-frequent letter, "brute-force" using predecessor / successor
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## Thank you!



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