

Combining Initial Segments of Lists

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Problem

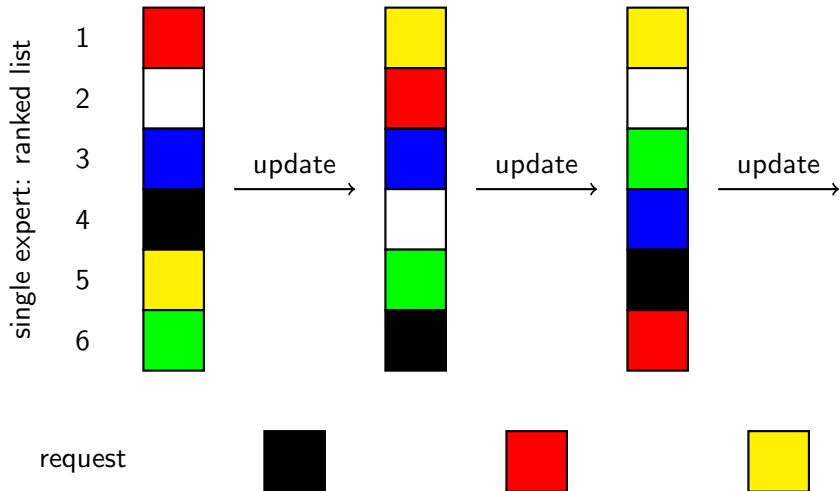
We want to help the user choose a color

We have access to **intelligent palettes** (our “experts”)

- gray levels
- pastels
- “Web colors”
- flags of the world
- copper tones
- ...

and we want a **master algorithm** to combine their advice.

Intelligent palette example: flags of the world



Combining palettes

Not *one* but *several* intelligent palettes

4 palettes



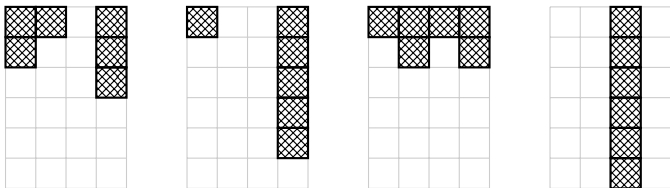
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A **combined palette** consists of 6 slots from tops of “expert” palettes

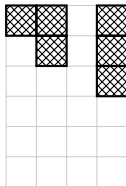


Combining palettes

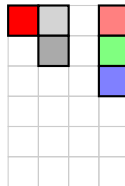
4 palettes



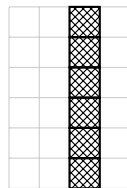
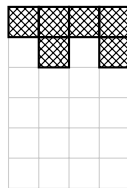
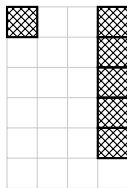
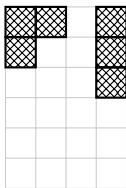
combined palette



result

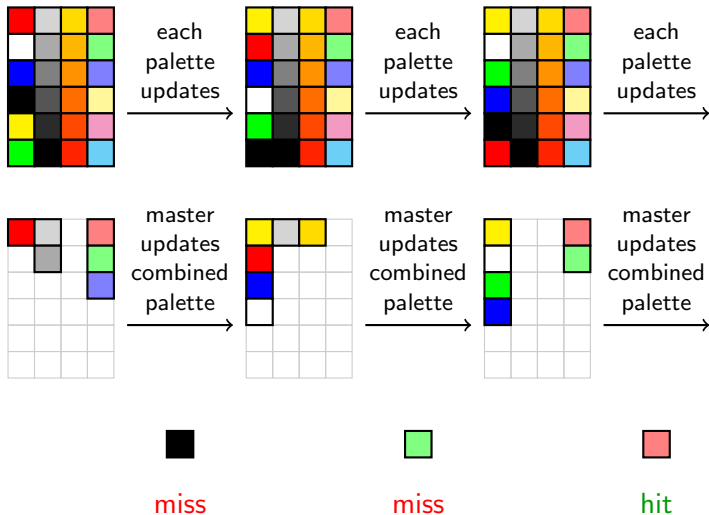


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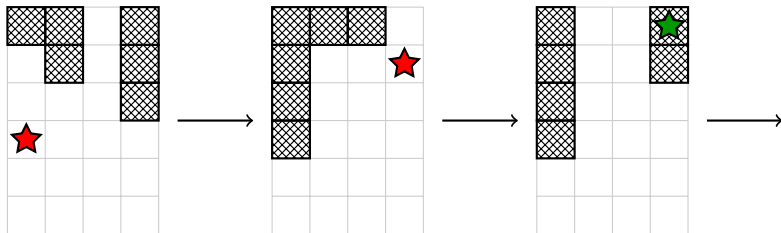
Master algorithm

A master algorithm chooses a combined palette each round.



A good master

A master algorithm chooses a combined palette each round.



Loss A combined palette incurs loss if it misses the requested item.

Goal Small regret compared to the best fixed combined palette.

(regret := # misses of master - # misses of best combined palette)

Abstract Problem

Setting We have K base lists of N slots each. Each round

- The base lists reveal their content.
- We select N items by taking initial segments of base lists.
- An item is requested. We either *hit* or *miss* it.

Goal Small regret w.r.t. the best fixed *combined list* in hindsight.

Difficulty There are $\binom{N+K-1}{K-1} \approx \left(\frac{N}{K}\right)^K$ such combined lists.

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N	K	combined lists
10	20	10^7
100	20	$5 \cdot 10^{21}$

Summary of Results I

Offline problem Finding the best combined list in hindsight is

- Easy: $O(KN^2)$ without duplicate items.
- Hard: decision problem is *NP complete* with duplicates.
(Is there a combined list with $\leq m$ misses for given items/requests?)

Relaxed problem: replace miss *indicator* with miss *count*

Not even close. (That's why it works.)

Online implementation Simulate (randomized) Hedge algorithm in $O(K N \log N)$ time per trial using Fast Fourier Transform.

Faster than offline!

Summary of Results II

Regret The expected regret of Hedge is at most

$$\sqrt{BK \ln \frac{N}{K}},$$

where B is the loss of the best combined list.

Lower bound Any algorithm has regret at least

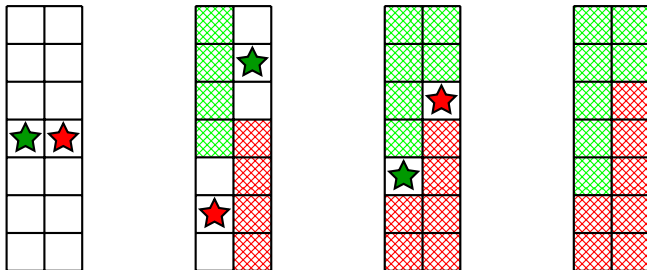
$$\sqrt{\max \left\{ \underbrace{B \ln N}_{K \text{ missing}}, \underbrace{BK}_{\ln \frac{N}{K} \text{ missing}} \right\}}.$$

Open problem Gap between bounds

Lower bound flavour

$K = 2$ base lists. **Reduction** to series of 2-expert games.

Key every combined list must miss one of the star items.



- $S = \log_2(N + 1)$ many phases.
- each phase, enforce loss $B/S + \sqrt{\frac{B}{S\pi}}$.
- master loss is $B + \sqrt{B \log_2(N + 1)/\pi}$.
- best combined list has loss B .

Conclusion

- Combining initial segments of lists
- Without duplicates:
 - Efficient implementation of Hedge algorithm
 - Faster than offline
 - Fast Fourier Transform
- With duplicates:
 - Hardness result
 - Transition to miss counts (cf attribute errors)
- Regret bound
- Two complementary* lower bounds