# A Compilation Based Approach to Finding Centroids and Minimum Covering States in Planning 

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## Motivation

- Suppose we have a set of possible goals
- One of these goals will "arrive" later, but we now have time to prepare for it
- We should go to either:
- a centroid state - one that minimizes the average distance to each possible goal
- a minimum covering state - one that minimizes the maximum distance to each possible goal
- Problem was first presented by Pozanco et. al. [PEFB19]


## Example



## Example



## Example



## Problem Setting

The setting here is STRIPS with multiple possible goals. Formally,
$\Pi=\langle F, A, I, \mathscr{G}, C\rangle$, where:

- $F$ is a set of facts describing the possible states of the world, $2^{F}$
- $A$ is a set of actions - each action $a \in A$ is $\langle p r e(a), \operatorname{add}(a), \operatorname{del}(a)\rangle$ with cost $C(a)$
- $I \subseteq F$ is the initial state of the world, and
- $\mathscr{G}$ is a set of possible goals, where each possible goal $G \in \mathscr{G}$ is a set of facts $G \subseteq F$. A state $s$ satisfies a goal if $G \subseteq s$


## Problem Objective

Denote by $h^{*}(s, G)$ the cost of an optimal path from state $s$ to a state $s^{\prime}$ such that $G \subseteq s^{\prime}$

- State $s$ is a centroid iff: $s$ is reachable from $I$, and $\sum_{i=1}^{n} h^{*}\left(s, G_{i}\right)$ is minimal (equivalent to minimizing average distance)
- State $s$ is a minimum covering state iff: $s$ is reachable from $I$, and $\max _{i=1}^{n} h^{*}\left(s, G_{i}\right)$ is minimal
The objective is to find either a centroid or a minimum covering state, possibly also optimizing over the cost to get there


## Inspiration

- The problem statement (and example) are very similar to finding worst case distinctiveness (wcd) in Goal Recognition Design (GRD) [KGK14]
- Reminder: the wcd is the maximal number of steps an agent can take from the initial state before its goal becomes clear
- Finding wcd is done via compilation to classical planning
- It turns out, the compilation for finding centroid states is very similar


## Compilation: Illustrated



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## Caveat: WCD $\neq$ Centroid $\neq$ Min-cover



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| $G$ |  |  | $C \cdots$ | $G$ |
| :---: | :---: | :---: | :---: | :---: |
|  |  |  | $C$ | $G$ |
|  |  | $M$ |  | $\vdots$ |
|  |  |  |  | $\vdots$ |
| $G$ |  |  |  | $\vdots$ |

## Centroid Compilation

Given $\Pi=\left\langle F, A, I, \mathscr{G}=\left\{G_{1}, \ldots G_{n}\right\}, C\right\rangle$ we define $\Pi^{\prime}=\left\langle F^{\prime}, A^{\prime}, I^{\prime}, G^{\prime}, C^{\prime}\right\rangle$, where:

- $F^{\prime}=\left\{f_{i} \mid f \in F, i=1 \ldots n\right\} \cup\{$ split, unsplit $\}$,
- $A^{\prime}=\left\{a_{i} \mid a \in A, i=1 \ldots n\right\} \cup\left\{a_{t} \mid a \in A\right\} \cup\{$ do-split $\}$, where
- $a_{t}$ is the together version of action $a$, affecting all of the $f_{i}$ facts, and is possible only before splitting
- $a_{i}$ is the separate version of action a for goal $i$, affecting only the $f_{i}$ variables, and is only possible after splitting
- The do-split action allows the agents to split
- $I^{\prime}=\left\{f_{i} \mid f \in I, i=1 \ldots n\right\} \cup\{$ unsplit $\}$
- $G^{\prime}=\left\{f_{i} \mid f \in G_{i}, i=1 \ldots n\right\}$


## Centroid Compilation vs. wcd Compilation

The only difference between the wcd compilation and this compilation are the costs:

- In wcd, we want to maximize the costs of the "together" actions, so the costs are
- $C\left(a_{t}\right)=n C(a)-\varepsilon$
- $C\left(a_{i}\right)=C(a)$
- In finding centroids, we only care about the costs of the "separate" actions, so the costs are
- $C\left(a_{t}\right)=0$
- If we want the compilation to find an optimal path to the centroid, we can set $C\left(a_{t}\right)=\varepsilon$ for a small enough $\varepsilon$
- $C\left(a_{i}\right)=C(a)$
- In all cases $C$ (do-split) $=0$


## Centroid Compilation: Theoretical Results

## Theorem

An optimal solution for $\Pi^{\prime}$ gives us a centroid state for the original task П.

## Proof sketch.

The compilation finds paths from the initial state to all goals. The cost of a plan for the compilation is the sum of costs after splitting, thus the state where it splits is a centroid.

## Centroid Compilation: Optimizations

- We can force the agents to act in order after splitting - first agent 1 (until it reaches its goal), then agent $2, \ldots$.
- This reduces permutations of essentially the same plans


## Finding Minimum Covering States

- Unfortunately, the max operator in minimum covering states is not additive
- Thus, we do not have a compilation which directly finds a minimum covering state in the general case
- We present a compilation which, given some cost budget $B$, checks whether there is some reachable state $s$ such that the maximum cost of reaching any possible goal $G_{i} \in \mathscr{G}$ from $s$ is at most $B$
- An binary search over $B$ will find minimum covering states (starting by doubling $B$ until the compilation is solvable)
- This is similar to the compilation for finding the wcd with non-optimal agents with deception budget [KGK15]


## Minimum Covering Compilation: Version 1 (numeric)

The compilation is the same as the centroid compilation, except

- We add $n$ new numerical variables, $B_{1} \ldots B_{n}$
- The value of $B_{i}$ in the initial state is 0
- $B_{i}<B-C\left(a_{i}\right)$ is added to pre $\left(a_{i}\right)$, and $B_{i}+=C\left(a_{i}\right)$ to the effects of $a_{i}$
- Note that we only care about the cost of reaching the goals after splitting, so $a_{t}$ actions are unmodified


## Theorem

Let $\Pi^{\prime}$ be a numerical planning task with budget $B$ as described above. Then $\Pi^{\prime}$ is solvable iff there exists some reachable state s such that $\max _{G \in \mathscr{G}}, h^{*}(s, G) \leq B$.

## Minimum Covering Compilation: Version 2 (unit cost actions)

- If all actions are unit cost, we can compile finding the minimum covering state to classical planning (without binary search)
- After splitting agents take turns executing actions in a round robin manner (without the optimization for enforcing the order between the agents)
- The compilation is implemented by:
- Adding $n$ new facts, turn ${ }_{i}$ for $i=1 \ldots n$
- For each $a_{i}$ action, we add turn ${ }_{i}$ to $\operatorname{pre}\left(a_{i}\right)$, turn ${ }_{i+1} \bmod n$ to $\operatorname{add}\left(a_{i}\right)$, and turn ${ }_{i}$ to $\operatorname{del}\left(a_{i}\right)$
- Adding NOOP actions - one for each agent, to allow agents to wait after reaching their goal
- The costs actions are 1 for actions of agent 1 after splitting, 0 for all others (agent 1 is guaranteed to act in every round)


## Empirical Evaluation

- We compared our compilation (C) to the exhaustive search approach (E) presented in the previous work
- Used several IPC domains and grid navigation with X\% obstacles
- Underlying planner was the same in both cases: Fast Downward [Hel06] with $\mathrm{A}^{*}$ [HNR68] and the Imcut heuristic [HD09]
- Time limit of 1 hour, memory limit of 16GB


## Empirical Results

|  | Centroid |  |  |  | Minimum Covering |  |  |  |
| :--- | ---: | ---: | ---: | ---: | ---: | ---: | ---: | :---: |
| Domain | C | E | Spdup | Cd | Cb | E | Spdup |  |
| blocks-w | 10 | 10 | 41.25 | 10 | 10 | 10 | 7.10 |  |
| ferry | $\mathbf{1 0}$ | 0 | - | $\mathbf{1 0}$ | $\mathbf{1 0}$ | 0 | - |  |
| gripper | $\mathbf{1 0}$ | 2 | 741.59 | $\mathbf{1 0}$ | $\mathbf{1 0}$ | 2 | 749.91 |  |
| hanoi | $\mathbf{1 0}$ | 6 | 372.86 | $\mathbf{1 0}$ | $\mathbf{1 0}$ | 6 | 355.36 |  |
| logistics | $\mathbf{1 0}$ | 2 | 195.32 | $\mathbf{1 0}$ | $\mathbf{1 0}$ | 2 | 188.97 |  |
| IPC | $\mathbf{5 0}$ | 20 | 226.17 | $\mathbf{5 0}$ | $\mathbf{5 0}$ | 20 | 204.05 |  |
| grid 5\% | $\mathbf{1 0}$ | 7 | 56.16 | 0 | 0 | $\mathbf{7}$ | - |  |
| grid 10\% | $\mathbf{1 0}$ | 8 | 92.20 | 1 | 0 | $\mathbf{7}$ | 0.19 |  |
| grid 15\% | 10 | 10 | 93.16 | 0 | 1 | $\mathbf{1 0}$ | - |  |
| grid 20\% | $\mathbf{1 0}$ | 9 | 74.12 | 0 | 0 | $\mathbf{9}$ | - |  |
| grid | $\mathbf{4 0}$ | 34 | 80.27 | 1 | 1 | $\mathbf{3 3}$ | 0.19 |  |
| TOTAL | $\mathbf{9 0}$ | 54 | 134.31 | 51 | 51 | $\mathbf{5 3}$ | 194.34 |  |

## Empirical Results: Takeaways

- On IPC domains, compilation based approach is about 200X faster than baseline
- On Grid
- Finding centroids using compilation is 80 X faster
- Finding min cover states using compilation is much slower - due to the small size of the state space


## Conclusion

- We presented a compilation based approach to finding centroids and minimum covering states
- Empirical performance for centroids is state-of-the-art
- Empirical performance for minimum covering states varies


## Thank You

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## Questions?

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