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1. Dota 2

- Dota 2 is a Multiplayer Online Battle Arena (MOBA)
- 123 possible heroes in 2 teams of 5 -> 10¹⁴ possible combinations
- Many items to choose from
- Focus on the hero 'Slark' and the item 'Hurricane Pike'
- OpenDota API provides all match data

CAUSALINFERENCE ON DOTA 2

2. Front-Door Adjustment

- Causal inference method that uses a mediator to determine the causal effect of a treatment
- Nullifies confounders
- Front-Door Criterion:
 - 1. Mediator intercepts all paths from treatment to outcome
 - 2. No backdoor path from treatment to mediator
 - 3. All backdoor paths from mediator to outcome are blocked by treatment

4. Results

- Results are obtained by running the code ٠ from the manual implementation and the Ananke package
- Stelios Avgousti's (a group-mate) results were are also included

	Win	Hurricane Pike	Total
Slark Picked	3946	1682	8213
Slark Not Picked	117017	42728	233713
Total	120963	44410	241926

Table 1: Frequencies in the data-set

Method	ATE	95% CI
Manual implementation	-0.00246	[-0.00346, -0.00148]
Ananke (Primal-IPW)	0.0098	[0.0060, 0.0139]
Ananke (Dual-IPW)	0.1961	[0.1170, 0.2740]
Ananke (APIPW)	0.1961	[0.1153, 0.2694]
Ananke (Eff-APIPW)	0.1961	[0.1137, 0.2745]
Stelios' results	-0.01119	NA

Table 2: Average treatment effects

5. Discussion

- Results are very diverse, with different possible interpretations
- Manual implementation and Primal IPW:
 - Come close to the 'true' value from Stelios' results
 - OR are statistically insignificant
- Other IPW versions are very large, which • seems unlikely considering the number of heroes in the game and Slark's win-rate (48,0%)
- Results might be so different because: •
 - Simplified representation of the game
 - Overfitting on the data
 - Not suitable for this scenario
- Condition 1 not fully met since there are other mediating variables, example being other counter-items.

- Hurricane Pike'

Confounders

- scenario
- unmeasurable
- results



6. Conclusion

Front-door adjustment is not a very suitable method for determining the casual effects of picking a hero on the outcome of the game Different interpretations already suggest that there are no clear one-sided answers, which makes the method unreliable for this

The method might work better if the dataset was more representative of the game, but some things such as player skill level are

Applying front-door adjustment on other, simpler scenarios might give more accurate