# Possible Approach to March Madness Dataset

### **Overview**

To solve this problem, we will create 2 entitysets with different "views" of the data. We will perform feature engineering on each view and then merge the results together.

The first entityset will create features for a particular pair of teams that are matching. The second entityset will create features for an individual team.

### **Notes**

- This obviously doesn't follow the actual dataset, but hopefully you can see the parallel's between my schema and the actual data.
- In your work, you should rename entities and column names to make things more clear

## **Entity Set 1**

### matchups entity

This entity has a row for every unique pairing of teams irrespective of home or away

match_id	team 1	team 2
duke_unc	duke	unc
kentucky_louisville	kentucky	louisville

### matchup\_logs entity

this entity has the outcomes of the matchups that occured

game	match_id	team_1_win	team_2_win	team1points	team2points
game_id_1	duke_unc	True	False	100	98
game_id_2	duke_unc	False	True	88	108
game_id_3	kentucky_louisville	True	False	86	76
game_id_4	duke_unc	?	?	?	?

#### Notes:

- The entities are related by the <code>match\_id</code> column
- The target entity for DFS would be matchup logs
- The last row represents a match up we want to make a prediction for
- I excluded it, but we need to add a time\_index column for each game, so we can use cutoff times
- When we call DFS, we pass cutoff times for each instance in matchup\_logs where the time is the time of the game. This will make sure we only use data from previous games when calculating features
- When we run DFS, we will want to use cutoff times
- To create feature like "percent of time duke won against unc" add interesting values [True, False] to team\_1\_win, team\_2\_win variables

## **Entity Set 2**

## team entity

this entity has a row for each team



### team\_game\_logs entity

This entity contains the outcomes of games for a single team's point of view. This means each game will have

#### 2 rows for each team

id	game	team	opponent	side	team_points	opponent_points	win
1	game_id_1	duke	unc	Home	100	98	True
2	game_id_1	unc	duke	Away	98	100	False
3	game_id_2	duke	unc	Away	88	108	False
4	game_id_3	unc	duke	Home	108	88	True
5	game_id_3	louisville	kentucky	Home	76	86	False
6	game_id_3	kentucky	louisville	Away	86	76	True
7	game_id_4	duke	unc	Home	?	?	?
8	game_id_4	unc	duke	Away	?	?	?

#### Notes

- these entities are connected by a relationship between team columns
- the target entity will be team game logs
- I excluded it, but we need to add a time\_index column for each game, so we can use cutoff times
- When we call DFS, we pass cutoff times for each instance in team game logs where the time is the time of
  the game. This will make sure we only use data from previous games when calculating features
- The value for game id \* should match between entitysets
- To create feature like "average points when home team" add interesting values ['Home', 'Away'] to side

## Merging to get final matrix

After features are calculated for both entities, we can merge together based on

- 1. Merge output of ES 1 with output of ES2 on game, team\_1 / game, team
- 2. Then merge output of step 1 with output of ES2 on game, team 2 / game, team