**March Machine Learning Mania 2017**

Overview: To predict the outcomes of this year's US men's college basketball tournament using a combination of rich historical data and computing power, while the ground truth unfolds on national television.

In the first stage of the competition, Kagglers will rely on results of past tournaments to build and test models. We encourage you to post any useful external data as a dataset. In the second stage, competitors will forecast outcomes of all possible match-ups in the 2017 tournament. You don't need to participate in the first stage to enter the second. The first stage exists to incentivize model building and provide a means to score predictions. The real competition is forecasting the 2017 results.

Available Data:

1. Teams

This file identifies the different college teams present in the dataset. Each team has a 4 digit id number.

1. Seasons

This file identifies the different seasons included in the historical data, along with certain season-level properties.

* "season" - indicates the year in which the tournament was played
* "dayzero" - tells you the date corresponding to daynum=0 during that season. All game dates have been aligned upon a common scale so that the championship game of the final tournament is on daynum=154. Working backward, the national semifinals are always on daynum=152, the "play-in" games are on days 134/135, Selection Sunday is on day 132, and so on. All game data includes the day number in order to make it easier to perform date calculations. If you really want to know the exact date a game was played on, you can combine the game's "daynum" with the season's "dayzero". For instance, since day zero during the 2011-2012 season was 10/31/2011, if we know that the earliest regular season games that year were played on daynum=7, they were therefore played on 11/07/2011.
* "regionW/X/Y/Z" - by convention, the four regions in the final tournament are always named W, X, Y, and Z. Whichever region's name comes first alphabetically, that region will be Region W. And whichever Region plays against Region W in the national semifinals, that will be Region X. For the other two regions, whichever region's name comes first alphabetically, that region will be Region Y, and the other will be Region Z. This allows us to identify the regions and brackets in a standardized way in other files. For instance, during the 2012 tournament, the four regions were East, Midwest, South, and West. Being the first alphabetically, East becomes W. Since the East regional champion (Ohio State) played against the Midwest regional champion (Kansas) in the national semifinals, that makes Midwest be region X. For the other two (South and West), since South comes first alphabetically, that makes South Y and therefore West is Z. So for this season, the W/X/Y/Z are East,Midwest,South,West.

1. RegularSeasonCompactResults

This file identifies the game-by-game results for 32 seasons of historical data, from 1985 to 2015. Each year, it includes all games played from daynum 0 through 132 (which by definition is "Selection Sunday," the day that tournament pairings are announced). Each row in the file represents a single game played.

* "season" - this is the year of the associated entry in seasons.csv (the year in which the final tournament occurs)
* "daynum" - this integer always ranges from 0 to 132, and tells you what day the game was played on. It represents an offset from the "dayzero" date in the "seasons.csv" file. For example, the first game in the file was daynum=20. Combined with the fact from the "season.csv" file that day zero was 10/29/1984, that means the first game was played 20 days later, or 11/18/1984. There are no teams that ever played more than one game on a given date, so you can use this fact if you need a unique key. In order to accomplish this uniqueness, we had to adjust one game's date. In March 2008, the SEC postseason tournament had to reschedule one game (Georgia-Kentucky) to a subsequent day, so Georgia had to actually play two games on the same day. In order to enforce this uniqueness, we moved the game date for the Georgia-Kentucky game back to its original date.
* "wteam" - this identifies the id number of the team that won the game, as listed in the "teams.csv" file. No matter whether the game was won by the home team or visiting team, "wteam" always identifies the winning team.
* "wscore" - this identifies the number of points scored by the winning team.
* "lteam" - this identifies the id number of the team that lost the game.
* "lscore" - this identifies the number of points scored by the losing team.
* "numot" - this indicates the number of overtime periods in the game, an integer 0 or higher.
* "wloc" - this identifies the "location" of the winning team. If the winning team was the home team, this value will be "H". If the winning team was the visiting team, this value will be "A". If it was played on a neutral court, then this value will be "N". Sometimes it is unclear whether the site should be considered neutral, since it is near one team's home court, or even on their court during a tournament, but for this determination we have simply used the Kenneth Massey data in its current state, where the "@" sign is either listed with the winning team, the losing team, or neither team.

1. RegularSeasonDetailedResults

This file is a more detailed set of game results, covering seasons 2003-2016. This includes team-level total statistics for each game (total field goals attempted, offensive rebounds, etc.) The column names should be self-explanatory to basketball fans (as above, "w" or "l" refers to the winning or losing team):

* wfgm - field goals made
* wfga - field goals attempted
* wfgm3 - three pointers made
* wfga3 - three pointers attempted
* wftm - free throws made
* wfta - free throws attempted
* wor - offensive rebounds
* wdr - defensive rebounds
* wast - assists
* wto - turnovers
* wstl - steals
* wblk - blocks
* wpf - personal fouls

1. TourneyCompactResults

This file identifies the game-by-game NCAA tournament results for all seasons of historical data. The data is formatted exactly like the regular\_season\_compact\_results.csv data. Note that these games also include the play-in games (which always occurred on day 134/135) for those years that had play-in games.

1. TourneyDetailedResults

This file contains the more detailed results for tournament games from 2003 onward.

1. TourneySeeds

This file identifies the seeds for all teams in each NCAA tournament, for all seasons of historical data. Thus, there are between 64-68 rows for each year, depending on the bracket structure.

* "season" - the year
* "seed" - this is a 3/4-character identifier of the seed, where the first character is either W, X, Y, or Z (identifying the region the team was in) and the next two digits (either 01, 02, ..., 15, or 16) tells you the seed within the region. For play-in teams, there is a fourth character (a or b) to further distinguish the seeds, since teams that face each other in the play-in games will have the same first three characters. For example, the first record in the file is seed W01, which means we are looking at the #1 seed in the W region (which we can see from the "seasons.csv" file was the East region). This seed is also referenced in the "tourney\_slots.csv" file that tells us which bracket slots face which other bracket slots in which rounds.
* "team" - this identifies the id number of the team, as specified in the teams.csv file

1. TourneySlots

This file identifies the mechanism by which teams are paired against each other, depending upon their seeds. Because of the existence of play-in games for particular seed numbers, the pairings have small differences from year to year. If there were N teams in the tournament during a particular year, there were N-1 teams eliminated (leaving one champion) and therefore N-1 games played, as well as N-1 slots in the tournament bracket, and thus there will be N-1 records in this file for that season.

* "season" - the year
* "slot" - this uniquely identifies one of the tournament games. For play-in games, it is a three-character string identifying the seed fulfilled by the winning team, such as W16 or Z13. For regular tournament games, it is a four-character string, where the first two characters tell you which round the game is (R1, R2, R3, R4, R5, or R6) and the second two characters tell you the expected seed of the favored team. Thus the first row is R1W1, identifying the Round 1 game played in the W bracket, where the favored team is the 1 seed. As a further example, the R2W1 slot indicates the Round 2 game that would have the 1 seed from the W bracket, assuming that all favored teams have won up to that point. The slot names are different for the final two rounds, where R5WX identifies the national semifinal game between the winners of regions W and X, and R5YZ identifies the national semifinal game between the winners of regions Y and Z, and R6CH identifies the championship game. The "slot" value is used in other columns in order to represent the advancement and pairings of winners of previous games.
* "strongseed" - this indicates the expected stronger-seeded team that plays in this game. For Round 1 games, a team seed is identified in this column (as listed in the "seed" column in the tourney\_seeds.csv file), whereas for subsequent games, a slot is identified in this column. In the first record of this file (slot R1W1), we see that seed W01 is the "strongseed", which during the 1985 tournament would have been Georgetown. Whereas for games from Round 2 or later, rather than a team seed, we will see a "slot" referenced in this column. So in the 33rd record of this file (slot R2W1), it tells us that the winners of slots R1W1 and R1W8 will face each other in Round 2. Of course, in the last few games of the tournament - the national semifinals and finals - it's not really meaningful to talk about a "strong seed" or "weak seed", but those games are represented in the same format for the sake of uniformity.
* "weakseed" - this indicates the expected weaker-seeded team that plays in this game, assuming all favored teams have won so far. For Round 1 games, a team seed is identified in this column (as listed in the "seed" column in the tourney\_seeds.csv file), whereas for subsequent games, a slot is identified in this column.

**Phase 1: Planning (02/06)**

Methodology:

1. Leverage regular season results to predict the tournament match up results

*(Using this method may lose the seed information but you could have more data points in return. Another bias is the validation dataset. You test well on the regular season game results but what you truly need to predict is the tournament game result. Keep trying this method until this has proved to be not significant)*

1. Use historical tournament results but append regular season information. This may have fewer data points but less bias

Level of modeling dataset: ID\_SeasonName, the combination of season and team level

Target variable: Team 1 result (W/L)

Predictors:

* Seed (completed on 2/7):
  + team 1 seed
  + team 2 seed
  + seed differential
  + team 1 last year seed (if not in tournament then assume 20)
  + team 2 last year seed (if not in tournament then assume 20)
  + team 1 seed differential (this year – last year)
  + team 2 seed differential (this year – last year)
* Team regular season metrics:
  + Winning % (Games\_played, Games\_won)
  + Average scoring (Scores / Games\_played)
  + Average scores allowed (Scores\_opp / Games\_played)
  + Average scoring differential (Average scoring – Average scores allowed)
  + Number of overtime games (sum(numot))
  + Overtime game winning % ()
  + Location: H/A/N winning % ()
  + Field goal made (fgm / Games\_played)
  + Field goal attempted (fga / Games\_played)
  + Field goal percentage (field goal made / field goal attempted)
  + Field goal percentage allowed (sum(fgm\_opp) / sum(fga\_opp))
  + Field goal percentage differential (Field goal percentage – Field goal percentage allowed)
  + Three pointers made (sum(tpm) / Games\_played)
  + Three pointers attempted (sum(tpa) / Games\_played)
  + Three pointer percentage % (Three pointers made / Three pointers percentage)
  + Three pointers percentage allowed
  + Three pointer percentage differential
  + Free throws made
  + Free throws attempted
  + Free throws percentage %
  + Free throw % differential
  + Offensive rebounds
  + Offensive rebounds allowed
  + Offensive rebounds differential
  + Defensive rebounds
  + Defensive rebounds allowed
  + Defensive rebounds differential
  + Total rebounds
  + Total rebounds allowed
  + Total rebounds differential
  + Offensive rebounds / total rebounds
  + Assists
  + Assists allowed
  + Turnovers
  + Turnovers forced
  + Assists / turnovers
  + Assists / points per game
  + Steals
  + Steals / Turnovers
  + Blocks
  + Personal fouls
  + Personal fouls draw
  + Net rating per game
  + Last season appearance indicator
  + IE: (PTS + FGM + FTM - FGA - FTA + DREB + (.5 \* OREB) + AST + STL + (.5 \* BLK) - PF - TO) / (GmPTS + GmFGM + GmFTM - GmFGA - GmFTA + GmDREB + (.5 \* GmOREB) + GmAST + GmSTL + (.5 \* GmBLK) - GmPF - GmTO)
  + EFF: ((Points + Rebounds + Assists + Steals + Blocks) - ((Field Goals Att. - Field Goals Made) + (Free Throws Att. -Free Throws Made) + Turnovers))

*\*\* Comment: Additionally, build differentials between team1 and team2 where applicable in the modeling dataset*

* Additional dynamic metrics that are available (To be added on 02/10):
  + Lag winning %
  + Dynamic total tournament appearance since 1985
  + Historical tournament winning %
  + Historical tournament average scoring
  + Historical average scoring allowed
  + Historical average scoring differential
  + Historical overtime winning %
  + Close game winning % (assume <10 is a close game)
  + Top 25 winning % (<https://www2.isye.gatech.edu/~jsokol/lrmc/lrmc.sort3.php>)
  + Historical match up winning %
  + Days rested winning %
  + True shooting percentage
  + Offensive efficiency rating
  + <http://stats.nba.com/help/glossary/>
* Additional data sources that might be helpful
  + Conference information
  + Player level stats (e.g. bench players scoring %)
  + Injury reports (e.g. # starting five absence games)
  + Recruitment reports (e.g. # top 50 players recruited)
  + Dynamic ranking (e.g. ranking differential between regular season start and regular season end)

**Phase 2: Data Manipulation and Exploratory Analysis**

Goal: To prepare the final modeling datasets and detect strong predictors to the target variables

**Phase 3: Algorithm Solution Building**

\*\* Random Forest works best (Log loss 0.16, ranked 1st) with the methodology building models based on the tournament game-level results that using ~140 predictors from Team1 seed, Team2 seed, Differentials, regular season metrics (02/09)

Pitch to Internal Team:

Prediction level: Team 1 winning percentage over Team 2 for any possible match up. The prediction should be at the game level.

Data sources:

* Seed: historical seed information
* Box scores: historical detailed box scores for regular season / tournament from 2003-2016

Algorithm methodology: Train the model with only the tournament game information but added regular season stats as predictors

Data manipulation: rolled up to the game level with information of the two teams

Predictors (~180):

* Absolute value of Team 1 and Team 2
  + Seed\_num\_Team1
  + Score average diff Team1
  + Seed num Team2
  + Games won Team1
  + Winning percentage Team1
* Differentials
  + Seed differential
  + Score average diff diff
  + FGP3\_opp\_diff
  + Win Score diff (points+rebounds+steals+0.5\*Assists+0.5\*blocked shots-field goal attempts-turnovers-0.5\*FTA-0.5\*PF)

Model: Random Forest with ~40 top predictors

Model Evaluation: LogLoss function

Concerns:

1. Additional available data sources (player-level data, injury, recruitment, fans, etc.)
2. Variables creation to solve the bias and skew for ranking
3. Methodology, variables are highly correlated with each other, which makes GLM hard to implement, requires PCs and unpacking works
4. Game dynamics (real-time environment changing, morale, player match-ups, coach strategies, etc.)