Fantasy Football

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As a markov chain

At each gameweek, we have 11 players that will play. Let's assume that these are the only players we have control of (i.e. no subs).

There are N total players on the game (N~650), and given some conditions, are available to be bought.

We will start assuming that we must make one transfer or less each gameweek, unlike in Fantasy football where you can pay 4 points to do an extra.

On gameweek one, the state of our team is:

•
$$\underline{s_1} = \{s_{1j}\}_{j=1}^{j=11}$$

i.e. s_1 is an eleven-vector

Here, we let s_{ij} represent the player in the j^{th} player slot on gameweek i.

Let's fix s_1 :

$$s_1 \equiv \{s_{1_1}, s_{1_2}, ..., s_{1_{11}}\} \in \{1, 2, ..., N\}^{11}$$

- Note that we cannot select a single player more than once, and so we have for fixed i:
 - $\forall (j,k) \in \{1,2,...,11\}^2, s_{i_j} \neq s_{i_k} \text{ for } j \neq k.$

For gameweek 2, we can now transition to another state. Essentially, we have the following two choices:

- We could swap out one $s_{1_j} \in \{1, 2, ..., N\}$ for s_{2_j} .
- We could make no transfers so that $s_1 = s_2$
- We have $11(N-11) \sim 7029$ choices for transfer as we can swap any of the 11 players out, and for each player that we swap out, we have N-11 choices, as we cannuot swap for the same people.

We must filter out: - The players that are out of pay range, given who they are substituted for. - Therefore we will need to apply the filter for each player, and then bind the tables together.

Finding the possible transfers

We will first give a toy example, assuming we are in gameweek one. To do this we will use the {fplr} package. First we give ourselves a random team of 11 to begin with.

```
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
library(fplr)
# Create a dataframe of players and teams simple information
team_lookup <- fpl_get_teams() %>% select(name, id)
player_lookup <- fpl_get_player_all() %>%
  left_join(team_lookup, c("team" = "id")) %>%
  select(-team) %>%
 rename("team_name" = name) %>%
  select(first_name, second_name, team_name, id)
set.seed(1871)
s_1 <- fpl_get_player_all() %>%
  sample_n(11, replace = FALSE) %>%
 left_join(team_lookup, c("team" = "id")) %>%
  select(-team) %>%
 rename("team_name" = name) %>%
  select(first_name, second_name, team_name, id, everything())
s_1
## # A tibble: 11 x 67
##
      first_name second_name team_name
                                           id chance_of_playi~ chance_of_playi~
##
      <chr>
                 <chr>
                             <chr>>
                                        <int>
                                                         <int>
                                                                           <int>
   1 Shane
##
                 Long
                             Southamp~
                                          361
                                                            NA
                                                                              NA
## 2 Matt
                             Aston Vi~
                                           42
                                                            NA
                                                                              NA
                 Targett
## 3 Michael
                 Obafemi
                             Southamp~
                                          381
                                                            25
                                                                              25
                                           25
                                                                               0
## 4 Matteo
                 Guendouzi
                             Arsenal
                                                             0
## 5 Joël
                 Veltman
                             Brighton
                                           67
                                                            NA
                                                                              NA
## 6 Erik
                 Pieters
                                           83
                                                                             100
                             Burnley
                                                           100
## 7 Ellis
                 Simms
                             Everton
                                          596
                                                             0
                                                                               0
## 8 Danny
                 Ings
                             Southamp~
                                          366
                                                           100
                                                                             100
## 9 James
                 Trafford
                             Man City
                                          595
                                                            NA
                                                                              NA
## 10 Shane
                 Duffy
                             Brighton
                                          58
                                                             0
                                                                              0
## 11 John
                 Ruddy
                             Wolves
                                          453
                                                            NA
                                                                              NA
## # ... with 61 more variables: code <int>, cost_change_event <dbl>,
## #
       cost_change_event_fall <int>, cost_change_start <dbl>,
       cost change start fall <int>, dreamteam count <int>, element type <int>,
## #
       ep_next <dbl>, ep_this <dbl>, event_points <int>, form <dbl>,
## #
       in_dreamteam <lgl>, news <chr>, news_added <chr>, now_cost <dbl>,
```

```
## #
       photo <chr>, points_per_game <dbl>, selected_by_percent <dbl>,
## #
       special <lgl>, squad_number <lgl>, status <chr>, team_code <int>,
## #
       total points <int>, transfers in <int>, transfers in event <int>,
## #
       transfers_out <int>, transfers_out_event <int>, value_form <dbl>,
## #
       value_season <dbl>, web_name <chr>, minutes <int>, goals_scored <int>,
## #
       assists <int>, clean_sheets <int>, goals_conceded <int>, own_goals <int>,
       penalties_saved <int>, penalties_missed <int>, yellow_cards <int>,
## #
       red_cards <int>, saves <int>, bonus <int>, bps <int>, influence <dbl>,
## #
## #
       creativity <dbl>, threat <dbl>, ict_index <dbl>, influence_rank <int>,
## #
       influence_rank_type <int>, creativity_rank <int>,
       creativity_rank_type <int>, threat_rank <int>, threat_rank_type <int>,
## #
       ict_index_rank <int>, ict_index_rank_type <int>,
## #
       corners_and_indirect_freekicks_order <int>,
## #
       corners_and_indirect_freekicks_text <chr>, direct_freekicks_order <int>,
## #
       direct_freekicks_text <chr>, penalties_order <int>, penalties_text <chr>
```

Okay so we now have 11 players. Therefore the following transfers are:

```
library(purrr)

# Find players that do not have an id in s1 and then give current id as id_out. Then bind rows.

transfers <- s_1$id %>%
   map(function(id) {
    transf <- player_lookup %>%
        rename("id_in" = id) %>% # here id is not the argument to map, but the column in player_lookup filter(!(id_in %in% s_1$id)) %>%
        mutate("id_out" = id)
}) %>%
   bind_rows()
transfers
```

```
## # A tibble: 6,941 x 5
##
      first name
                                           team name id in id out
                     second name
##
                     <chr>>
                                           <chr>
                                                     <int> <int>
      <chr>>
                     Özil
                                                              361
## 1 Mesut
                                           Arsenal
## 2 Sokratis
                     Papastathopoulos
                                                         2
                                                              361
                                           Arsenal
## 3 David
                     Luiz Moreira Marinho Arsenal
                                                         3
                                                              361
## 4 Pierre-Emerick Aubameyang
                                           Arsenal
                                                         4
                                                              361
   5 Cédric
                                                         5
##
                     Soares
                                           Arsenal
                                                              361
                                                         6
## 6 Alexandre
                                                              361
                     Lacazette
                                           Arsenal
                                                         7
## 7 Shkodran
                     Mustafi
                                           Arsenal
                                                              361
## 8 Bernd
                     Leno
                                           Arsenal
                                                         8
                                                              361
## 9 Granit
                     Xhaka
                                           Arsenal
                                                         9
                                                              361
## 10 Pablo
                     Marí
                                           Arsenal
                                                        10
                                                              361
## # ... with 6,931 more rows
```

Note how we have 642 players to begin with, and thus should expect 11 * (642 - 11) total available transfer choices. This is what we have:

```
nrow(transfers)
```

```
## [1] 6941
```

Now let's give a cost to each transfer. Clearly, this is 'player in cost' - 'player out cost':

```
## # A tibble: 6,941 x 8
      transfer_cost first_name second_name team_name id_in id_out now_cost_in
##
##
              <dbl> <chr>
                               <chr>
                                            <chr>
                                                      <int>
                                                             <int>
                                                                          <dbl>
##
   1
              1.6
                    Mesut
                               Özil
                                            Arsenal
                                                          1
                                                               361
                                                                            6.7
##
   2
             -0.300 Sokratis
                               Papastatho~ Arsenal
                                                          2
                                                               361
                                                                            4.8
##
   3
              0.3
                    David
                               Luiz Morei~ Arsenal
                                                          3
                                                               361
                                                                           5.4
                    Pierre-Em~ Aubameyang Arsenal
##
   4
                                                          4
                                                               361
                                                                           11.3
              6.2
##
   5
             -0.5
                    Cédric
                               Soares
                                            Arsenal
                                                          5
                                                               361
                                                                           4.6
                                                                           8.3
##
   6
              3.2
                    Alexandre Lacazette
                                            Arsenal
                                                          6
                                                               361
##
   7
             -0.100 Shkodran
                               Mustafi
                                            Arsenal
                                                          7
                                                               361
                                                                            5
##
             -0.100 Bernd
                                                               361
                                                                            5
   8
                               Leno
                                            Arsenal
                                                          8
   9
                    Granit
                               Xhaka
                                                          9
                                                               361
                                                                            5.2
              0.1
                                            Arsenal
             -0.700 Pablo
                               Marí
                                                               361
                                                                            4.4
## 10
                                            Arsenal
                                                         10
## # ... with 6,931 more rows, and 1 more variable: now_cost_out <dbl>
```

Assume we already have a model for player's points

We now assume that we have modelled, elsewhere, the expected number of points a player will score in a particular gameweek. Number of points in gameweek i is given by: p_i

We give this expected number of points as