My title*

My subtitle if needed

First author

Another author

November 29, 2024

First sentence. Second sentence. Third sentence. Fourth sentence.

1 Introduction

^{*}Code and data are available at: https://github.com/RohanAlexander/starter_folder.

2 Data

```
cleaned data %>% select(!c(ball, innings, match id)) %>% head()
# A tibble: 6 x 17
  year venue
                     over batting_team bowling_team striker bowler runs_off_bat
  <dbl> <chr>
                   <dbl> <chr>
                                        <chr>
                                                     <chr>
                                                                           dbl>
                                                             <chr>>
1 2021 MA Chidamba~
                         1 Mumbai Indi~ Royal Chall~ RG Sha~ Moham~
                                                                               0
2 2021 MA Chidamba~
                         1 Mumbai Indi~ Royal Chall~ RG Sha~ Moham~
                                                                               0
3 2021 MA Chidamba~
                         1 Mumbai Indi~ Royal Chall~ RG Sha~ Moham~
                                                                               2
4 2021 MA Chidamba~
                         1 Mumbai Indi~ Royal Chall~ RG Sha~ Moham~
                                                                               0
5 2021 MA Chidamba~
                         1 Mumbai Indi~ Royal Chall~ RG Sha~ Moham~
                                                                               1
6 2021 MA Chidamba~
                         2 Mumbai Indi~ Royal Chall~ RG Sha~ KA Ja~
                                                                               1
# i 9 more variables: wickets_lost_yet <dbl>, wicket <lgl>, target <dbl>,
   run_rate <dbl>, batting_style <chr>, batter_playing_role <chr>,
    bowling_style <chr>, bowler_playing_role <chr>, prev_over_wickets <int>
```

2.1 Measurement

2.2 Predictor Variables

```
num_bowlers_per_type <- cleaned_data %>%
  group_by(bowling_style) %>%
  summarise(
    num_bowlers = n_distinct(bowler)
)

num_batters_per_type <- cleaned_data %>%
  group_by(batting_style) %>%
  summarise(
    num_bowlers = n_distinct(striker)
)
```

2.3 Relationship Between Wickets and other Variables

```
cleaned_data %>% head()
```

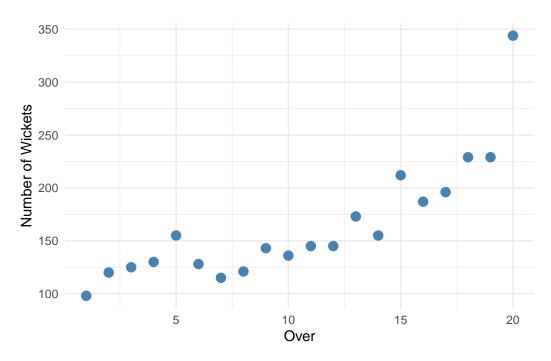
```
# A tibble: 6 x 20
 match_id year venue
                         innings over ball batting_team bowling_team striker
     <dbl> <dbl> <chr>
                            <dbl> <dbl> <dbl> <chr>
                                                           <chr>
                                                                        <chr>
1 1254058 2021 MA Chida~
                                1
                                      1
                                            2 Mumbai Indi~ Royal Chall~ RG Sha~
2 1254058 2021 MA Chida~
                                            3 Mumbai Indi~ Royal Chall~ RG Sha~
                                1
3 1254058 2021 MA Chida~
                                            4 Mumbai Indi~ Royal Chall~ RG Sha~
                                1
                                      1
4 1254058 2021 MA Chida~
                                1
                                      1
                                            5 Mumbai Indi~ Royal Chall~ RG Sha~
5 1254058 2021 MA Chida~
                                1
                                      1
                                            6 Mumbai Indi~ Royal Chall~ RG Sha~
6 1254058 2021 MA Chida~
                                1
                                      2
                                            1 Mumbai Indi~ Royal Chall~ RG Sha~
# i 11 more variables: bowler <chr>, runs_off_bat <dbl>,
  wickets_lost_yet <dbl>, wicket <lgl>, target <dbl>, run rate <dbl>,
   batting_style <chr>, batter_playing role <chr>, bowling_style <chr>,
   bowler_playing_role <chr>, prev_over_wickets <int>
stadium_boundaries <- cleaned_data %>%
  group_by(venue) %>%
 summarise(
   num_matches = n_distinct(match_id),
   num wickets = sum(wicket == TRUE),
 ) %>% arrange(desc(num_wickets), desc(num_matches))
ggplot(stadium_boundaries, aes(x = venue, y = (num_wickets/num_matches))) +
 geom_bar(stat = "identity", fill = "steelblue") +
 labs(
      x = "Stadium Name",
      y = "Wickets Per Match") +
  theme_minimal() +
```

coord_flip()



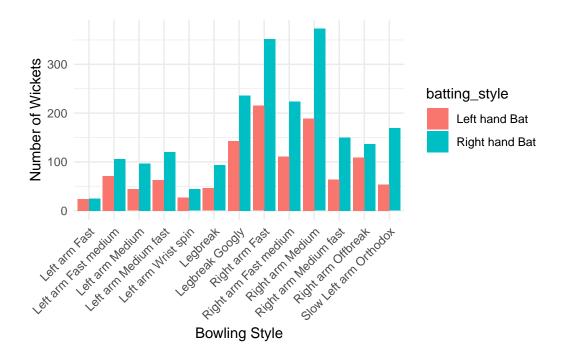
```
over_boundaries <- cleaned_data %>%
  group_by(over) %>%
  summarise(
    num_wickets = sum(wicket == TRUE),
    num_balls = n()
) %>% arrange(desc(num_wickets), desc(num_balls))

ggplot(over_boundaries, aes(x = over, y = num_wickets)) +
  geom_point(color = "steelblue", size = 3) +
  labs(
    x = "Over",
    y = "Number of Wickets") +
  theme_minimal()
```



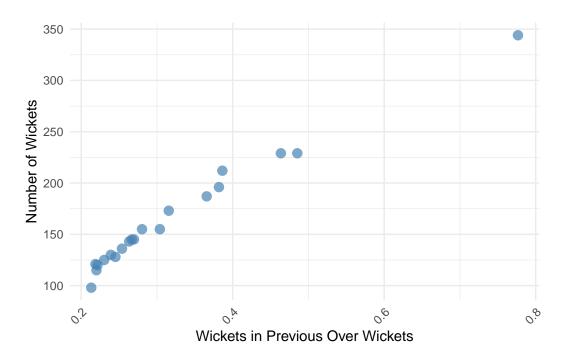
```
bowling_batting_matchup_boundaries <- cleaned_data %>%
  group_by(bowling_style, batting_style) %>%
  summarise(
   num_wickets = sum(wicket == TRUE),
   num_balls = n(),
) %>% arrange(desc(num_wickets), desc(num_balls))
```

`summarise()` has grouped output by 'bowling_style'. You can override using the `.groups` argument.



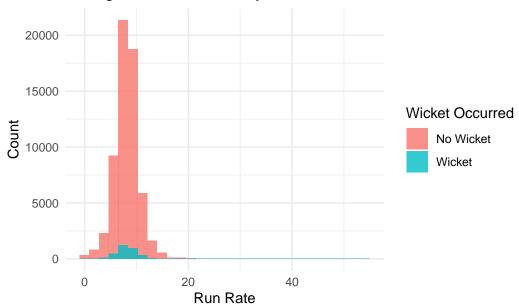
```
wickets_prev_over_wickets <- cleaned_data %>%
  group_by(over) %>%
  summarise(
    num_wickets = sum(wicket == TRUE),
    prev_over_wickets = mean(prev_over_wickets),
    num_balls = n(),
  ) %>% arrange(desc(num_wickets), desc(num_balls))

ggplot(wickets_prev_over_wickets, aes(x = prev_over_wickets, y = num_wickets)) +
    geom_point(color = "steelblue", size = 3, alpha = 0.7) +
    labs(
        x = "Wickets in Previous Over Wickets",
        y = "Number of Wickets") +
    theme_minimal() +
    theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



```
ggplot(cleaned_data, aes(x = run_rate, fill = factor(wicket))) +
  geom_histogram(position = "identity", alpha = 0.8, bins = 30) +
  labs(
    x = "Run Rate",
    y = "Count",
    title = "Histogram of Run Rate by Wicket Occurrence"
  ) +
  scale_fill_discrete(name = "Wicket Occurred", labels = c("No Wicket", "Wicket")) +
  theme_minimal()
```

Histogram of Run Rate by Wicket Occurrence



3 Model

3.1 Model set-up

3.1.1 Model justification

4 Results

```
simple_glm_wicket_model <- readRDS(here("models/simple_glm_wicket_model.rds"))
#summary(simple_glm_wicket_model)
modelsummary(simple_glm_wicket_model)</pre>
```

```
complex_glm_wicket_model <- readRDS(here("models/complex_glm_wicket_model.rds"))
#summary(complex_glm_wicket_model)
modelsummary(complex_glm_wicket_model)</pre>
```

	(1)
(Intercept)	-3.569
	(0.048)
over	0.058
	(0.004)
Num.Obs.	51 587
AIC	20589.7
BIC	20607.4
Log.Lik.	-10292.858
RMSE	0.22

	(1)
(Intercept)	-4.139
	(0.052)
over	0.005
	(0.004)
prev_over_wickets	1.852
	(0.031)
Num.Obs.	51587
AIC	16590.1
BIC	16616.7
Log.Lik.	-8292.074
RMSE	0.21

```
overly_complex_glm_wicket_model <- readRDS(here("models/overly_complex_glm_wicket_model.rds")
#summary(overly_complex_glm_wicket_model)
modelsummary(overly_complex_glm_wicket_model)</pre>
```

5 Simple Model Summary

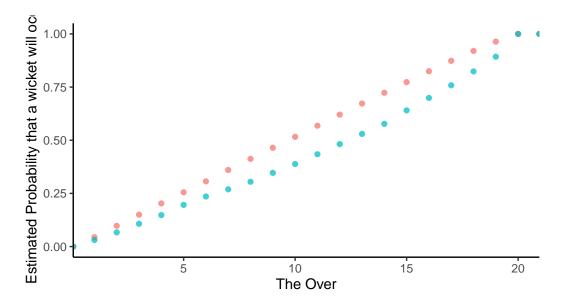
```
overly_complex_glm_wicket_model <- complex_glm_wicket_model <- readRDS(here("models/overly_complex_glm_wicket_model)
#summary(overly_complex_glm_wicket_model)
modelsummary(overly_complex_glm_wicket_model)</pre>
```

```
simple_glm_wicket_model_predictions <-
predictions(simple_glm_wicket_model) |>
as_tibble()

simple_glm_wicket_model_predictions |>
mutate(wicket = factor(wicket)) |>
ggplot(aes(x = over, y = estimate, color = wicket)) +
stat_ecdf(geom = "point", alpha = 0.75) +
labs(
    x = "The Over",
    y = "Estimated Probability that a wicket will occur",
    color = "Was actually a wicket"
) +
theme_classic() +
theme(legend.position = "bottom")
```

	(1)
(Intercept)	$\frac{-4.316}{-4.316}$
1 /	(0.192)
over	0.005
	(0.004)
prev_over_wickets	1.853
	(0.031)
batting_styleRight hand Bat	0.044
	(0.045)
bowling_styleLeft arm Fast medium	0.099
	(0.207)
bowling_styleLeft arm Medium	0.255
	(0.212)
bowling_styleLeft arm Medium fast	0.205
	(0.206)
bowling_styleLeft arm Wrist spin	0.158
	(0.235)
bowling_styleLegbreak	0.257
	(0.212)
bowling_styleLegbreak Googly	0.207
	(0.196)
bowling_styleRight arm Fast	0.190
	(0.192)
$bowling_styleRight~arm~Fast~medium$	0.211
	(0.197)
$bowling_styleRight~arm~Medium$	0.109
	(0.193)
$bowling_styleRight~arm~Medium~fast$	0.165
	(0.204)
$bowling_styleRight~arm~Offbreak$	0.104
	(0.200)
$bowling_styleSlow\ Left\ arm\ Orthodox$	0.011
	(0.202)
Num.Obs.	51 587
AIC	16605.8
BIC 11	16747.4
Log.Lik.	-8286.914
RMSE	0.21

		(1)
(Intercept)		(1)
(Intercept)		-4.316
		(0.192)
over		0.005
_		(0.004)
prev_over_wicket	ts	1.853
		(0.031)
batting_styleRigh	nt hand Bat	0.044
		(0.045)
bowling_styleLeft	t arm Fast medium	0.099
		(0.207)
bowling_styleLeft	t arm Medium	0.255
		(0.212)
bowling_styleLeft	t arm Medium fast	0.205
		(0.206)
bowling_styleLeft	t arm Wrist spin	0.158
		(0.235)
bowling_styleLeg	break	0.257
		(0.212)
bowling_styleLeg	break Googly	0.207
		(0.196)
bowling_styleRig	ht arm Fast	0.190
		(0.192)
bowling_styleRig	ht arm Fast medium	0.211
		(0.197)
bowling_styleRig	ht arm Medium	0.109
		(0.193)
bowling_styleRig	ht arm Medium fast	0.165
		(0.204)
bowling_styleRig	ht arm Offbreak	0.104
		(0.200)
bowling styleSlov	w Left arm Orthodox	0.011
<u> </u>		(0.202)
Num.Obs.		51 587
AIC		16 605.8
BIC	12	16 747.4
Log.Lik.	± =	-8286.914
RMSE		0.21
101/1DT		0.21



Was actually a wicket • FALSE • TRUE

```
test_data_simple <- test_data
predictions <- predict(simple_glm_wicket_model, newdata = test_data_simple, type = "response

test_data_simple$predicted_wicket_prob <- predictions
test_data_simple <- test_data_simple %>%
    mutate(predicted_wicket = predicted_wicket_prob >= 0.5) %>%
    mutate(correct_prediction = predicted_wicket == wicket)

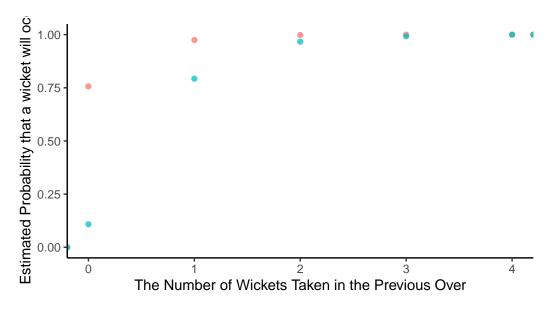
summary_results <- test_data_simple %>% group_by(wicket) %>%
    summarise(
    correct = sum(correct_prediction),
    incorrect = sum(!correct_prediction)
)

summary_results
```

5.1 Complex Model Summary

```
complex_glm_wicket_model_predictions <-
   predictions(complex_glm_wicket_model) |>
   as_tibble()

complex_glm_wicket_model_predictions |>
   mutate(wicket = factor(wicket)) |>
   ggplot(aes(x = prev_over_wickets, y = estimate, color = wicket)) +
   stat_ecdf(geom = "point", alpha = 0.75) +
   labs(
        x = "The Number of Wickets Taken in the Previous Over",
        y = "Estimated Probability that a wicket will occur",
        color = "Was actually a wicket"
   ) +
   theme_classic() +
   theme(legend.position = "bottom")
```



Was actually a wicket • FALSE • TRUE

```
test_data_complex <- test_data
predictions <- predict(complex_glm_wicket_model, newdata = test_data_complex, type = "respons
test_data_simple$predicted_wicket_prob <- predictions
test_data_simple <- test_data_simple %>%
```

```
mutate(predicted_wicket = predicted_wicket_prob >= 0.5) %>%
mutate(correct_prediction = predicted_wicket == wicket)

summary_results <- test_data_simple %>% group_by(wicket) %>%
summarise(
correct = sum(correct_prediction),
incorrect = sum(!correct_prediction)
)

summary_results
```

6 Discussion

6.1 First discussion point

If my paper were 10 pages, then should be be at least 2.5 pages. The discussion is a chance to show off what you know and what you learnt from all this.

6.2 Second discussion point

Please don't use these as sub-heading labels - change them to be what your point actually is.

6.3 Third discussion point

6.4 Weaknesses and next steps

Weaknesses and next steps should also be included.

Appendix

A Additional data details

```
bowling_batting_role_matchup_boundaries <- cleaned_data %>%
  group_by(bowling_style, batter_playing_role) %>%
  summarise(
   num_wickets = sum(wicket == TRUE),
   num_balls = n(),
) %>% arrange(bowling_style, batter_playing_role)
```

`summarise()` has grouped output by 'bowling_style'. You can override using the `.groups` argument.

bowling_batting_role_matchup_boundaries

```
# A tibble: 117 x 4
           bowling_style [13]
# Groups:
  bowling_style
                        batter_playing_role num_wickets num_balls
  <chr>
                        <chr>>
                                                   <int>
                                                             <int>
1 Left arm Fast
                        Allrounder
                                                      10
                                                               115
2 Left arm Fast
                        Batter
                                                       2
                                                                95
                                                       2
3 Left arm Fast
                        Batting Allrounder
                                                                79
4 Left arm Fast
                        Bowler
                                                       4
                                                                37
5 Left arm Fast
                        Bowling Allrounder
                                                       2
                                                                18
6 Left arm Fast
                        Middle order Batter
                                                       4
                                                                89
7 Left arm Fast
                        Opening Batter
                                                       4
                                                               174
                                                       9
8 Left arm Fast
                        Top order Batter
                                                               137
9 Left arm Fast
                        Wicketkeeper Batter
                                                      12
                                                               190
10 Left arm Fast medium Allrounder
                                                      25
                                                               503
# i 107 more rows
```

B Model details

B.1 Posterior predictive check

C References

What to cite: - cricketdata - ESPNCricinfo - Cricsheet - All tidyverse packages used