

Monash University: Assessment Cover Sheet

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Unit name	FIT5147 Data exploration and visualisation - S2 2021		
Lecturer's name	SARAH GOODWIN	Tutor's name	ANGEL DAS
Assignment name	Data Exploration Project Submission (33%)		Group Assignment: No Note, each student must attach a coversheet
Lab/Tute Class:	TUTORIAL 4	Lab/Tute Time:	12:00-14:00
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		Extension granted	<input type="checkbox"/>

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FIT5147 – Data Exploration and Visualization

Data Exploration Project – Report

Topic: The Influence of Anime

Done by

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Student ID: 31386709

Course: Master of Data Science

Introduction

In the past few years, Anime has become a phenomenon across the world [1]. Anime (animation made in Japan) has started to influence people worldwide. Being influenced by it myself, I wanted to dive into the market share of Anime. By doing so, we will come to know how Anime has grown over the years to the point where it has become a global sensation. In addition, we will be discovering the factors that make an anime series successful.

Data Wrangling

We will start of our adventure by getting our datasets from Kaggle named "MyAnimeList Dataset" [2] which is a collection of metadata and information about various Anime as well as its viewers. For this project we will be taking the raw version. we will take 3 datasets [2] ("UserList.csv", "UserAnimeList.csv" and "AnimeList.csv") which contains the data about the Viewers, Anime seen by the Viewers and Anime series respectively. The total size of the 3 datasets combined is 4.69 GB! So it is necessary to perform data wrangling or else we may not be able to visualize our data due to the issues caused by its Volume (For Instance, slow performance).

We start of our wrangling with the elephant in the room "UserAnimeList.csv" which is a massive 4.6 GB file that has all the data about the anime that each user has seen. So it is no surprise that it is so large. For wrangling we will be using R over tableau for Quality of life purposes. When we open the dataset we can see that there are 80 million records with 11 attributes.



```
UAL <- read_csv("UserAnimeList.csv")
head(UAL)
```

A tibble: 6 x 11

username	anime_id	my_watched_episodes	my_start_date	my_finish_date	my_score	my_status
<chr>	<dbl>	<dbl>	<date>	<date>	<dbl>	<dbl>
karthiga	21	586	<NA>	<NA>	9	1
karthiga	59	26	<NA>	<NA>	7	2
karthiga	74	26	<NA>	<NA>	7	2
karthiga	120	26	<NA>	<NA>	7	2
karthiga	178	26	<NA>	<NA>	7	2
karthiga	210	161	<NA>	<NA>	7	2

6 rows | 1-7 of 11 columns

Figure 1: A glimpse at our initial Data frame

We remove the unwanted columns as it removes a significant portion of the dataset. The only columns that are required by us is username, anime_id, my_watched_episodes, my_score, my_status.



```
{r}
names(UAL)

select(UAL,
  username,
  anime_id,
  my_watched_episodes,
  my_score,
  my_status)
```

Figure 2: The attributes/columns of UserAnimeList.csv

We then remove the other attributes from the data frame as a result we will only have the attributes we require.

```

{r}
head(UAL)

```

username	anime_id	my_watched_episodes	my_score	my_status
karthiga	21	586	9	1
karthiga	59	26	7	2
karthiga	74	26	7	2
karthiga	120	26	7	2
karthiga	178	26	7	2
karthiga	210	161	7	2

6 rows

Figure 3: The Wrangled UserAnimeList.csv

Before proceeding, we must look into what the values in my_status mean. The image below gives us an idea of it.

my_status in animelists tables contains integer values. This is their semantics:

- 1: watching
- 2: completed
- 3: on hold
- 4: dropped
- 6: plan to watch

other values are not known.

Figure 4: Legend for my_status column in UserAnimeList.csv

The data that is relevant to our analysis are with status 1,2,3 or 4. We do not have to consider the records with status 6 as planning to see does not guarantee them seeing it. We will remove records that are not of the status 1,2,3 or 4.

```

{r}
unique(UAL$my_status)

[1] 1 2 3 6 4 0 5 33 55

{r}
UAL <- UAL[UAL$my_status %in% c(1,2,3,4),]

{r}
unique(UAL$my_status)

[1] 1 2 3 4

```

Figure 5: Filtering based on my_status

Now we will move on to the next data set "AnimeList.csv". Looking at the dataset there seem we only require a few attributes only. As we did with the previous data set we will remove them.

anime_id	title	title_english	title_japanese	title_synonyms	image_url	type	source	episodes	status
1	11013 Inu x Boku SS	Inu x Boku Secret Service	犬と秘密	Yoku x Boku SS	https://myanimelist.co...	TV	Manga	12	Fin
2	2154 Seta no Hanayome	My Bride is a Mermaid	嫁神の夜姫	The Inland Sea Bride	https://myanimelist.co...	TV	Manga	26	Fin
3	5262 Shugo Chara! Doki	Shugo Chara! Doki	しゅごキャラ! どきっ	Shugo Chara Ninime, Shugo Chara: Second Year	https://myanimelist.co...	TV	Manga	51	Fin
4	721 Princess Tutu	Princess Tutu	プリンセスチュチュ	N/A	https://myanimelist.co...	TV	Original	38	Fin
5	12368 Bakuman, 3rd Season	Bakuman	バクマン!	Bakuman Season 3	https://myanimelist.co...	TV	Manga	25	Fin
6	6586 Yume no Ritsuka	N/A	夢色パティシエール	Yumeko Patissiere, Yumefit, Dream-Colored Pastry Chef, Y...	https://myanimelist.co...	TV	Manga	50	Fin
7	176 Uta no Prince	Uta no Prince	歌の王子様	N/A	https://myanimelist.co...	TV	Manga	26	Fin
8	2787 Shugaun no Shana II (Second)	Shugaun no Shana Season II	灼眼のシャナII-Second-	Shugaun no Shana 2	https://myanimelist.co...	TV	Light novel	24	Fin
9	4477 Nodame Cantabile: Paris-hen	N/A	のだめカンタービレ パリ篇	Nodame Cantabile Paris Chapter: Nodame Cantabile Paris-hen	https://myanimelist.co...	TV	Manga	11	Fin
10	853 Ouran Koukou Host Club	Ouran High School Host Club	桜蘭高校ホスト部	Ouran Koukou Host Club, Ouran Koukou Hostclub, Ouran Ko...	https://myanimelist.co...	TV	Manga	26	Fin
11	4414 Junjo Romantica 2	Junjo Romantica 2	純情ロマンチス2	Junjo Romantica Second Season, Junjo Yu Romantica II	https://myanimelist.co...	TV	Manga	12	Fin
12	7054 Kichou wa Maid-sama!	Maid Sama!	会長はメイド様!	Class President is a Maid!	https://myanimelist.co...	TV	Manga	26	Fin
13	11123 Seikai Hatsukoi - Words & Music's Greatest First Love 2	Seikai Hatsukoi 2	世界一初恋2	Seikai Hatsukoi 2, Seikai Hatsukoi 2	https://myanimelist.co...	TV	Manga	12	Fin
14	14227 Toran no Kabutsubun	My Little Monster	とらんの怪獣くん	Toran no Kabutsubun, The Monster Next Door, My Neigbo...	https://myanimelist.co...	TV	Manga	13	Fin
15	289 Beach	Beach	ビーチ	N/A	https://myanimelist.co...	TV	Manga	566	Fin
16	59 Chobits	Chobits	ちょびっツ	N/A	https://myanimelist.co...	TV	Manga	26	Fin
17	6546 Kimi ni Todoke	Kimi ni Todoke From Me to You	君に届け	Reaching You	https://myanimelist.co...	TV	Manga	25	Fin
18	1735 Naruto Shippuden	Naruto Shippuden	ナルト 疾風伝	Naruto Hurricane Chronicles	https://myanimelist.co...	TV	Manga	500	Fin
19	210 Ramen 1/2	Ramen 1/2	ラーメン1/2	Ramen 1/2, Ramen 1/2, Nattou Hen	https://myanimelist.co...	TV	Manga	161	Fin
20	4234 Toradora!	Toradora!	とらドラ!	Tiger X Dragon	https://myanimelist.co...	TV	Light novel	25	Fin
21	10020 Bakuman, 2nd Season	Bakuman	バクマン! 2ndシーズン	Bakuman Season 2	https://myanimelist.co...	TV	Manga	25	Fin
22	74 Gakuen Alice	Gakuen Alice	学園アリス	Campus Alice, Alice Academy	https://myanimelist.co...	TV	Manga	26	Fin
23	4722 Skip Beat!	Skip Beat!	スキップ・ビート!	N/A	https://myanimelist.co...	TV	Manga	25	Fin
24	14387 Ohayayo! 2	N/A	おはよう2	Ohayayo! 2	https://myanimelist.co...	TV	Manga	25	Fin
25	1357 Shounen Onmyoji	Shounen Onmyoji	少年陰陽師	The Young Saint Master, Teenager Onmyoji, Shounen Onmyo...	https://myanimelist.co...	TV	Light novel	26	Fin
26	10800 Ohayayo!	Ohayayo!	おはよう	Ohayayo!	https://myanimelist.co...	TV	Manga	25	Fin
27	3731 Natu ra Kiss	Natu ra Kiss	イナズマキス	Naughty Kiss, Teasing Kiss, Mischievous Kiss, Natuana Kiss...	https://myanimelist.co...	TV	Manga	25	Fin
28	9513 Beecubus	Beecubus	ハチキュウ	N/A	https://myanimelist.co...	TV	Manga	60	Fin
29	5838 Harassment Sensei-bun	Harassment Sensei-bun	電撃ハラスメント	Harassment Senseibun	https://myanimelist.co...	TV	Manga	39	Fin
30	9603 SKET Dance	SKET Dance	スケッチ・ダンス	N/A	https://myanimelist.co...	TV	Manga	77	Fin
31	7817 B-gata Hae!	Yamada-kun's First Time & Gata Hae!	12歳! 恋	N/A	https://myanimelist.co...	TV	4-koma manga	12	Fin
32	946 Cryer Shin-chan	Shin Chan	クレヨンしんちゃん	N/A	https://myanimelist.co...	TV	Manga	0	Cu
33	120 Fruits Basket	Fruits Basket	フルーツバスケット	Furuda	https://myanimelist.co...	TV	Manga	26	Fin
34	957 Seikoku Monogatari	The Story of Seikoku	海物語伝説	The Tale of Seikou Country, Story of the Land of Many-Colo...	https://myanimelist.co...	TV	Light novel	39	Fin

Figure 6: Overview of AnimeList.csv

```
{r}
names(anime)
...

[1] "anime_id"      "title"          "title_english"  "title_japanese" "title_synonyms" "image_url"
[7] "type"          "source"         "episodes"       "status"         "aired"         "aired_string"
[13] "aired"         "duration"       "rating"         "score"         "scored_by"     "rank"
[19] "popularity"    "members"        "favorites"      "background"     "premiered"     "broadcast"
[25] "related"       "producer"       "licensor"       "studio"         "genre"         "opening_theme"
[31] "ending_theme"
```

Figure 7: Columns in AnimeList.csv before wrangling

colnames(anime)					
[1]	"anime_id"	"title"	"title_english"	"type"	"source"
[7]	"status"	"duration"	"rating"	"score"	"studio"
					"episodes"
					"genre"

Figure 8: Columns in AnimeList.csv after Wrangling

We will now go to the last dataset “UserList.csv”. As usual, we will we will see the attributes present in it.

username	user_id	user_watching	user_completed	user_onhold	user_dropped	user_plantowatch	user_days_spent_watching	gender	location	birth_date	access_rank	join_date	last_online
karthiga	2255153	3	49	1	0	0	55.31	Female	Chennai, India	1990-04-29	N/A	2013-03-03	2014-02-04 01:32:00
RedvelvetDaisuki	1897806	61	396	39	0	206	118.07	Female	Manila	1995-01-01	N/A	2012-12-13	1900-05-10 02:47:00
Damianahui	37326	45	195	27	25	59	83.70	Male	Detroit Michigan	1991-08-01	N/A	2008-02-13	1900-03-24 12:48:00
bskal	228342	25	414	2	5	11	167.16	Male	Nayarit, Mexico	1990-12-14	N/A	2009-08-31	2014-05-12 16:35:00
shuzabille	2347781	36	72	16	2	25	35.48	N/A	N/A	N/A	2013-03-25	2015-09-09 21:54:00	
terume_jumaki	327331	5	5	0	0	0	15.20	Female	Malaysia, Kuantan	1998-06-24	N/A	2010-05-10	2012-10-18 19:06:00
Bas_C	5015094	35	114	6	20	175	30.81	Male	Nijmegen, Nederland	1999-10-24	N/A	2015-11-26	1900-05-10 13:53:00
punane	392500	1	0	0	0	0	2.97	N/A	N/A	N/A	2008-02-24	2011-02-13 02:42:00	
otaku4name	4583478	0	4	0	0	1	6.42	N/A	N/A	N/A	2015-04-27	2015-05-04 19:05:00	
sprite1989	102436	2583	102	0	0	22	30.86	Male	N/A	1989-07-29	N/A	2008-10-06	2017-10-30 05:26:00
thereedude	28328	11	703	27	29	79	206.68	Male	California	N/A	2007-12-29	2017-09-10 19:04:00	
Heihouka	12489	8	69	0	8	0	112.73	Male	N/A	N/A	2007-09-06	2015-08-30 04:36:00	
Himekita	3129315	2	87	2	0	28	20.12	Female	Poland	1996-09-26	N/A	2013-09-08	1900-04-27 10:52:00
Skallington	326733	0	3	0	0	0	4.32	Male	Norberg Sweden	1991-03-15	N/A	2010-05-09	2010-05-12 06:29:00
Slimak	61677	79	224	0	3	84	126.17	Male	Poland	1988-02-21	N/A	2008-05-18	1900-05-01 05:04:00
Elysun	5927342	7	4	0	0	4	2.09	Male	Naborska	1984-11-16	N/A	2016-12-25	2017-03-11 21:37:00
jemothy000	47167	6	0	0	0	2	7.71	Male	England, Manchester	1995-04-25	N/A	2008-04-01	2008-04-27 16:00:00
MixButterfly	2485327	66	3923	115	0	368	614.96	Female	~ Hungary ~	1992-01-16	N/A	2013-04-25	2018-05-17 13:31:00
BackAngel	15865	3	20	0	0	5	21.34	Male	N/A	1991-08-14	N/A	2007-10-03	2017-05-31 15:52:00
Not_Riad	4960716	1	99	5	2	14	52.78	N/A	Where even am i	N/A	2015-11-07	1900-05-10 17:56:00	
ProperBritish	253613	6	233	7	9	62	44.63	Male	United Kingdom	1992-02-20	N/A	2009-11-01	1900-04-19 03:47:00
Cranberry_cakes	32590	0	4	1	0	1	4.77	N/A	N/A	N/A	2008-01-20	2008-01-24 22:23:00	
ven_zakuu	211766	15	90	13	0	37	73.65	Female	manila	N/A	2009-07-22	2010-05-12 02:00:00	
konel	144049	11	433	5	2	0	166.24	Male	home	1988-01-16	N/A	2009-01-29	2017-05-31 05:57:00
ILLMATICc	2282031	47	80	10	14	151	61.82	Female	N/A	N/A	2013-03-09	1900-01-07 20:41:00	
Xmrl	1	4	230	8	89	60	133.32	Male	California	1985-03-04	N/A	2004-11-05	2017-05-16 03:20:00
vachachaseux	118593	7	31	18	0	29	23.25	Female	social-i	N/A	2008-11-28	2011-09-06 12:53:00	
heleynpdy	2529849	9	184	6	8	39	41.15	Female	Mexico	1999-08-30	N/A	2013-05-06	1900-05-13 15:25:00
cfordodd	61291	5	392	31	0	288	151.61	N/A	Sydney Australia N.S.W	N/A	2008-05-17	1900-04-18 00:38:00	
hopid_rnd	4538087	1	314	0	1	160	80.32	Male	Hamilton Ontario	1997-10-31	N/A	2015-04-04	1900-04-21 19:44:00
wetstange	452735	3	432	7	28	10	134.48	Female	Nevada USA	N/A	2011-02-23	1900-05-12 02:35:00	
dewsteddow311	3309	3	29	5	2	20	24.58	Female	Chicago, IL	1989-09-14	N/A	2007-06-06	2009-01-16 16:48:00
thasuboket	15867	5	146	15	25	15	63.02	Male	San Jose, CA	1984-09-08	N/A	2007-10-26	2015-08-28 01:17:00
YumYun_Pachina	270169	27	270	16	78	240	66.97	Female	N/A	N/A	2009-12-15	1900-05-11 18:17:00	
mon33500	189574	4	37	0	5	4	11.82	Female	N/A	N/A	2009-06-02	2014-11-10 16:00:00	
ThelFalken	340873	22	655	2	3	212	131.89	Male	somewhere really hot	1993-05-29	N/A	2010-06-13	1900-05-16 03:20:00
LCantierDN	14658	5	0	0	1	0	3.65	Male	Curess	1995-05-29	N/A	2007-09-23	2008-04-24 16:01:00
zZimhoZz	6951406	10	65	3	8	22	73.30	Male	Camagaribe, Pernambuco	2000-04-04	N/A	2018-02-03	2018-05-17 10:41:00
xXxTou94xXx	38148	0	0	0	0	0	0.00	N/A	N/A	N/A	2008-09-24	2008-09-24 16:29:00	
SaberioffKane	5844149	21	183	46	18	190	44.45	N/A	N/A	N/A	2016-11-17	2018-05-16 19:20:20	
glittermilk	502599	15	35	2	6	70	17.20	Female	austria-U+3002>	1994-04-13	N/A	2011-06-23	2012-08-03 06:47:00
aksum	60375	2	386	1	14	125	156.31	Male	N/A	N/A	2008-05-14	2017-08-03 06:14:00	
misumini	844589	11	116	16	1	101	21.77	N/A	N/A	N/A	2011-10-30	2017-08-26 05:54:00	
fruffydewolf	1537661	6	121	11	12	175	27.63	Female	Canada	1989-01-01	N/A	2012-09-24	2016-06-22 13:18:00

Figure 9: An overview of UserList.csv

colnames(users)			
[1]	"username"	"user_id"	"user_watching"
[5]	"user_onhold"	"user_dropped"	"user_plantowatch"
[9]	"gender"	"location"	"birth_date"
[13]	"join_date"	"last_online"	"stats_mean_score"
[17]	"stats_episodes"		"stats_rewatched"

Figure 10: The attributes of UserList.csv before wrangling

Out all the columns we need only need 3 columns. We will remove the other attributes as well as records with no location (location = NA) from the dataset.

```
users <- users[is.na(users$location) == F,]
users
```

A tibble: 156,796 x 3

username <chr>	gender <chr>	location <chr>
karthiga	Female	Chennai, India
RedvelvetDaisuki	Female	Manila
Damonashu	Male	Detroit, Michigan
bskai	Male	Nayarit, Mexico
terune_uzumaki	Female	Malaysia, Kuantan
Bas_G	Male	Nijmegen, Nederland
thetreedude	Male	California
HimeAria	Female	Poland
Skallington	Male	Norberg Sweden

Figure 11: UserList.csv after filtering

After we perform our Data Checking which will be explained in the next section we will merge all the tables which will form our final dataset.

Data Checking

All the data sets are skimmed through in order to spot and rectify any anomalies. Let us start with UserAnimeList.csv. We can see a few records where the value of my_watched_episodes is 0. This does not make any sense. So we will remove them and the my_watched_episodes column as it has served its purpose.

```
UAL[UAL$my_watched_episodes > 0, c("username", "anime_id", "my_score")]
```

A tibble: 57,934,010 x 3

username <chr>	anime_id <dbl>	my_score <dbl>
karthiga	21	9
karthiga	59	7
karthiga	74	7
karthiga	120	7
karthiga	178	7
karthiga	210	7
karthiga	232	6

Figure 12: The clean version of UserAnimeList.csv

We save the clean version of this file as a new csv called UserAnimeListFiltered.csv. Now we will check the next dataset AnimeList.csv here we can see 2 main errors. The episode duration is in the form “XX mins per episode” which isn’t suitable for visualisation. This will cause issues when we need to visualize based on the episode duration. So we create a user defined function called extract_mins that will return only the integer value. We will also change the column name to Duration_in_Minutes.

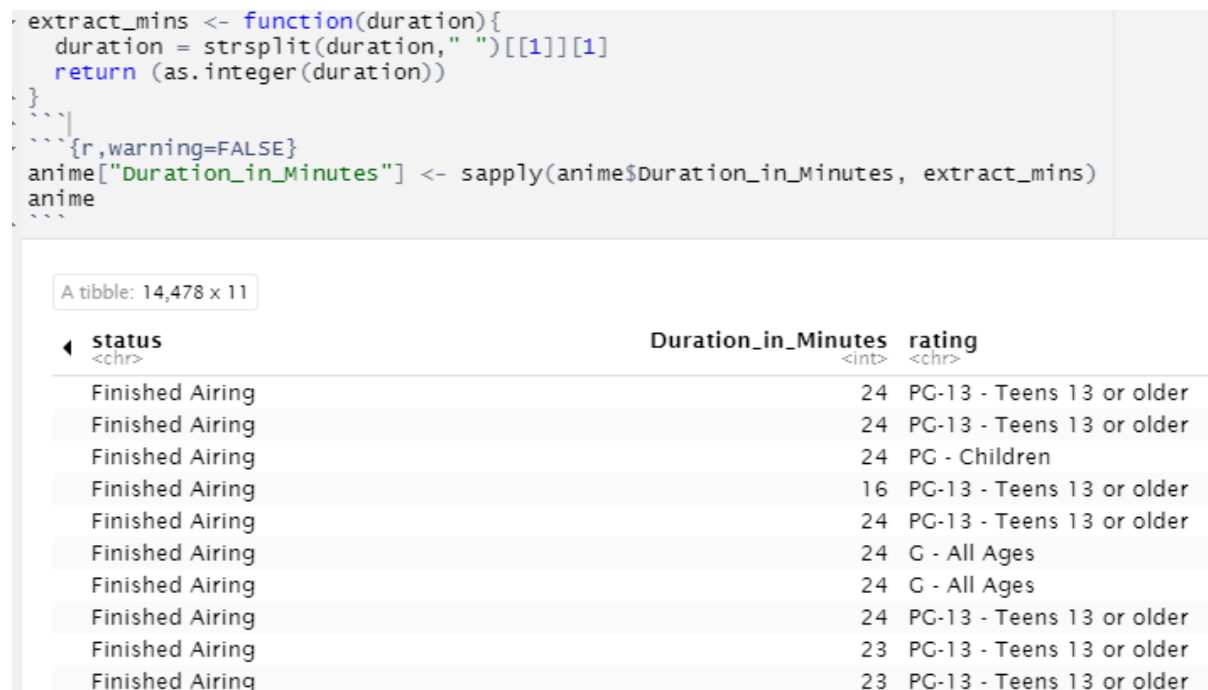


Figure 13: Type conversion of Episode Duration

Next is the genre which is multivalued will be really hard for us to visualize based on genre. We need to turn all the genres to boolean attributes assign the appropriate TRUE/FALSE value to it.



Figure 14: Assigning appropriate Boolean values

After removing the genre column, we save the cleaned dataset as a new csv called AnimeListFiltered.csv.


Now to the UsersList.csv dataset. The main issue is that the location is not in a tableau friendly format. In order for us to use the spatial data, the library countrycode [3] was used to convert the location. Once converted, we will remove the records where conversion wasn't possible. We then save it as a new csv called UsersFiltered.csv

username <chr>	gender <chr>	location <chr>	Location_Name <chr>
karthiga	Female	IND	India
bskai	Male	MEX	Mexico
terune_uzumaki	Female	MYS	Malaysia
HimeAria	Female	POL	Poland
Skallington	Male	SWE	Sweden
Slimak	Male	POL	Poland
MistButterfly	Female	HUN	Hungary
ProperBritish	Male	GBR	United Kingdom
helenply	Female	MEX	Mexico
cfoordddd	NA	AUS	Australia

Figure 15: The Cleaned UserList.csv

Now we need to an inner join between the datasets UsersFiltered.csv and UserAnimeListFiltered.csv. We will join the resulting data frame with the AnimeListFiltered.csv data set in tableau itself as performing an inner join increases the size of the data set to 7 GB.

UALxUser.csv is made of 2 tables. ⓘ



Sort fields: Data source order ▾ Show aliases Show hidden fields 1											
ime id (Ani...	Abc	Abc	Abc	Abc	#	Abc	#	Abc	#	Abc	Abc
imeListFiltered.csv	AnimeListFiltered.csv	AnimeListFiltered.csv	AnimeListFilt...	AnimeListFilt...	AnimeListFilter...	AnimeListFiltered...	AnimeListFiltered.csv	AnimeListFiltered.csv	AnimeListF...	AnimeListFiltered.csv	Studio
ime id (Ani...	Title	Title English	Type	Source	Episodes	Status	Duration in Mi...	Rating	score	Studio	
34547	Shoukoku no Altair	Altair: A Record ...	TV	Manga	24	Finished Airing	24	PG-13 - Teens 13 ...	7.55000	MAPPA	
250	Konjiki no Gash B...	Zatch Bell!	TV	Manga	150	Finished Airing	23	PG-13 - Teens 13 ...	7.64000	Toei Animation	
32603	Okusama ga Seit...	My Wife is the St...	TV	Manga	12	Finished Airing	8	R+ - Mild Nudity	6.81000	Seven	
29575	Mankitsu Happe...	NA	OVA	Visual novel	4	Finished Airing	30	Rx - Hentai	7.88000	Collaboration W...	
9201	Air Gear: Kuro no...	NA	OVA	Manga	3	Finished Airing	28	R+ - Mild Nudity	7.73000	Satelight	
2889	Bleach Movie 2: ...	Bleach the Movie...	Movie	Manga	1	Finished Airing	1	PG-13 - Teens 13 ...	7.56000	Studio Pierrot	
9289	Hanasaku Iroha	Hanasaku Iroha: ...	TV	Original	26	Finished Airing	24	PG-13 - Teens 13 ...	8.01000	P.A. Works	
9989	Ano Hi Mita Han...	anohana: The Flo...	TV	Original	11	Finished Airing	22	PG-13 - Teens 13 ...	8.55000	A-1 Pictures	
387	Haibane Renmei	Haibane Renmei	TV	Other	13	Finished Airing	24	PG-13 - Teens 13 ...	8.04000	Radix	
5958	Sora no Otoshim...	Heaven's Lost Pr...	TV	Manga	13	Finished Airing	24	PG-13 - Teens 13 ...	7.57000	AIC A.S.T.A.	
6512	Doatshuk	Doatshuk	TV	Manga	60	Finished Airing	24	PG-13 - Teens 13 ...	8.03000	Diogenes Blue	

Figure 16: The final Dataset.

Data Exploration

Now we will proceed with the visualisations in Tableau. We will start by analysing the market/viewership share of Anime. As anime is a visual medium, we can find its market by checking it viewership or TRP. We will be analysing it in the following 6 ways.

- **User Distribution:** the number of anime fans country wise.

- **Anime Distribution:** Tells us about the no of anime series that have caught the attention of the users. This will be done genre wise as well as overall.
- **Anime Viewership:** Gives us information about the overall TRP (total views overall) of anime in a country.
- **Most Viewed anime:** Gives us information about the most viewed anime in a particular country.
- **Highest rated anime:** Gives us the highest rated anime for each country. This is calculated by taking the average scores of each anime seen in the country and choosing the anime with the maximum average.
- **Anime's key of success:** we will be plot the scores of various anime series against factors like studio, episode length, Rating, Type and Genre.

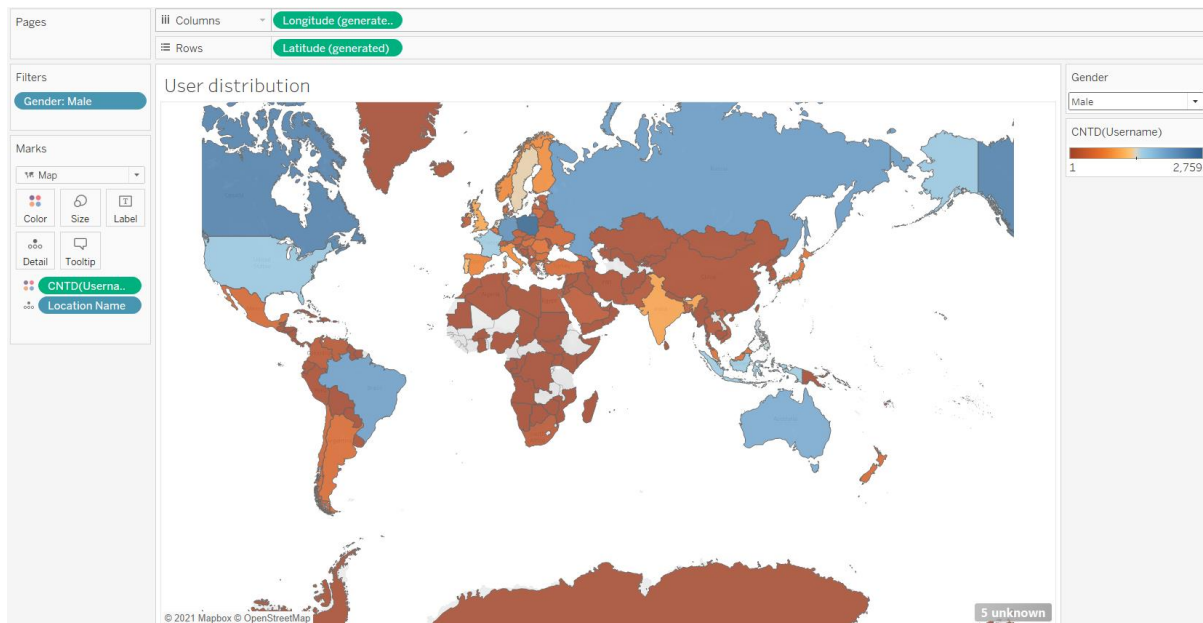


Figure 17: The number of Anime fans around the world

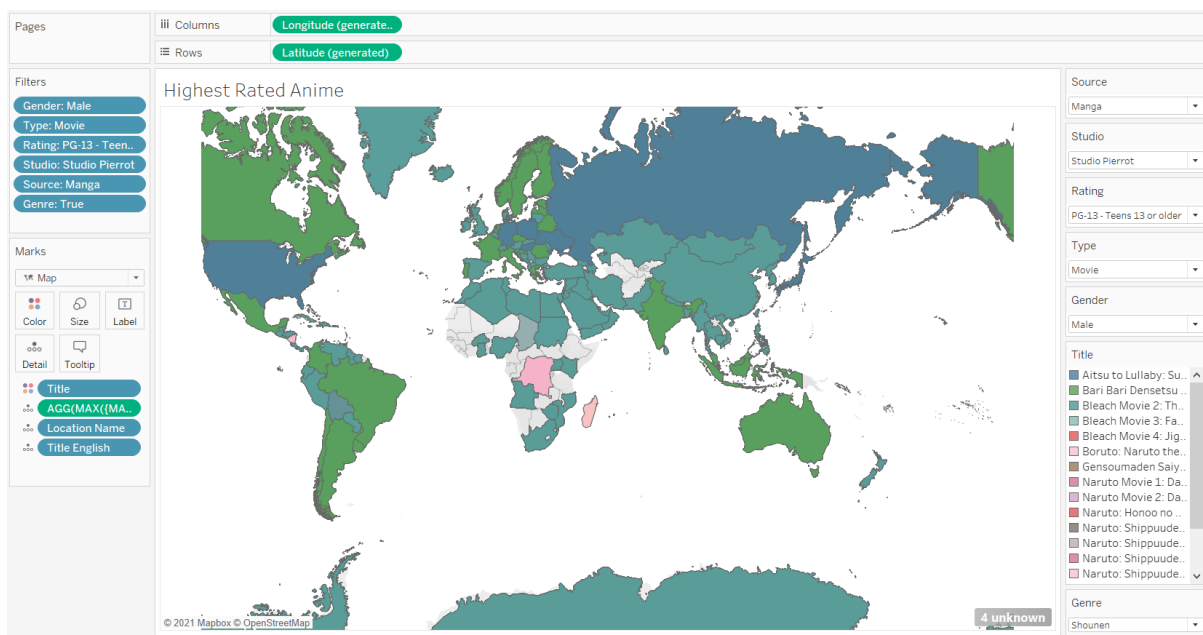


Figure 18: The visualisation of the Highest rated anime series

All the visualisations except the anime's keys of success will use spatial data. In every visualisation there will be the following filters

- **Gender:** We will have an option to filter our visualisation by choosing a gender. There are 4 unique types of genders Male, Female, Non-Binary(Transgender) and NA.
- **Type:** Type is the form in which it was released. The major types are Movie, TV, OVA (Original Video Animation), ONA (Original Net Animation), Special and Unknown.
- **Studio:** The Studio that made the anime. A few instances are Studio Bones, MAPPA, etc.
- **Rating:** Rating gives us the age restrictions given to an anime series. For Example, G, PG, etc.
- **Source:** It tells us the kind of source material from the anime and its characters are based on. Manga and Light Novels are examples of it.

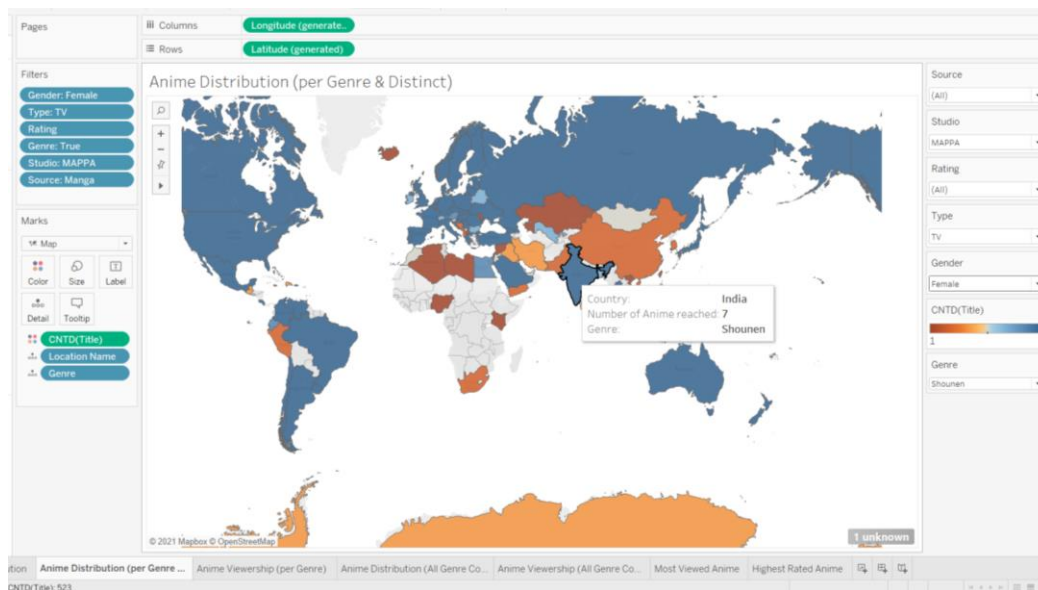


Figure 19: The anime Reach based on many factors

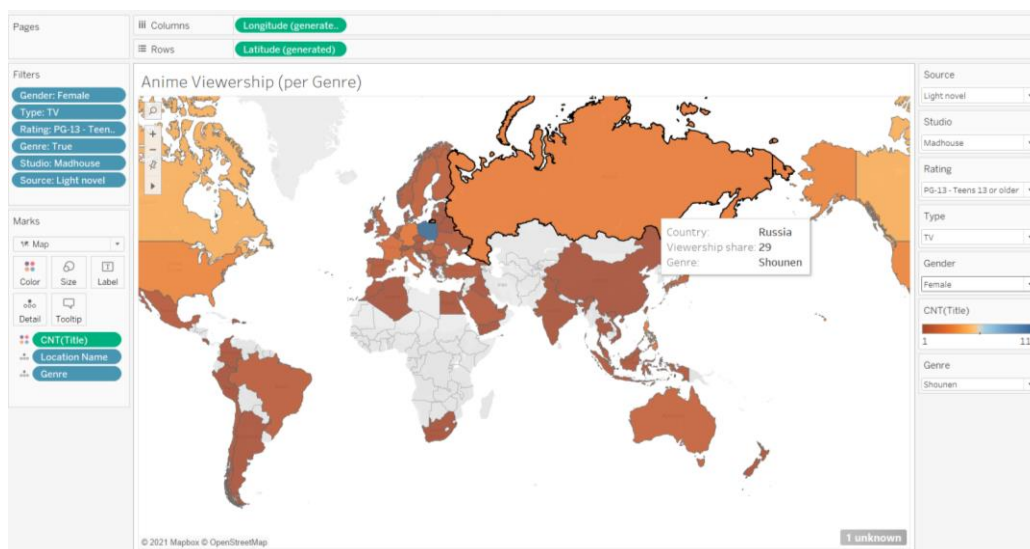


Figure 20: The Viewership Share of Anime based on many factors

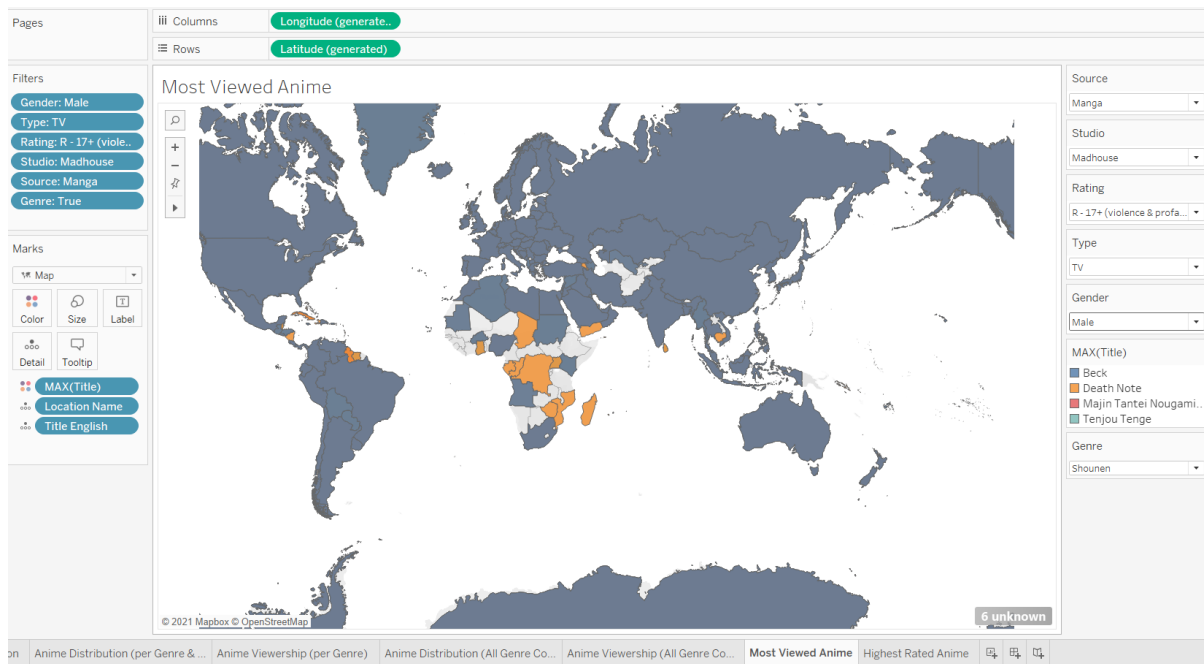


Figure 21: The most Viewed Country based on the filters

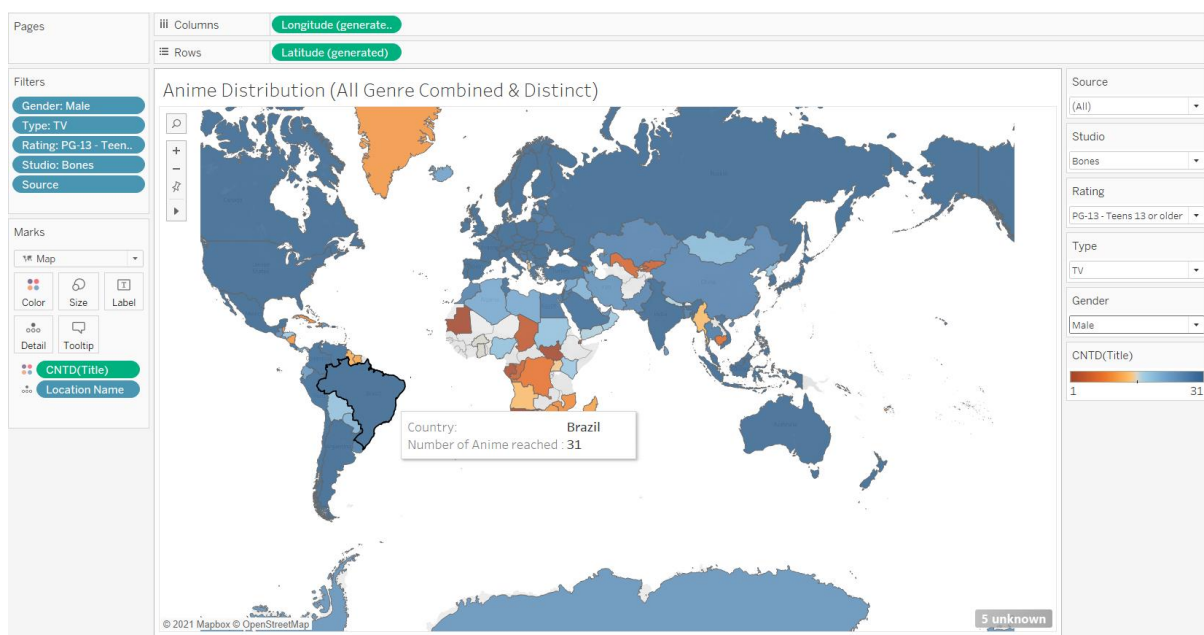


Figure 22: Overall Anime Distribution

By visualising we are able to find out the how different types of anime are received around the world. This will help us understand the global reach of the genre as well as the sub genres that are preferred by the community. Now we understand the tastes of the community and will give animation studios an idea before they produce another anime. We can also see the factors that will lead to an anime's success and how they affect them.

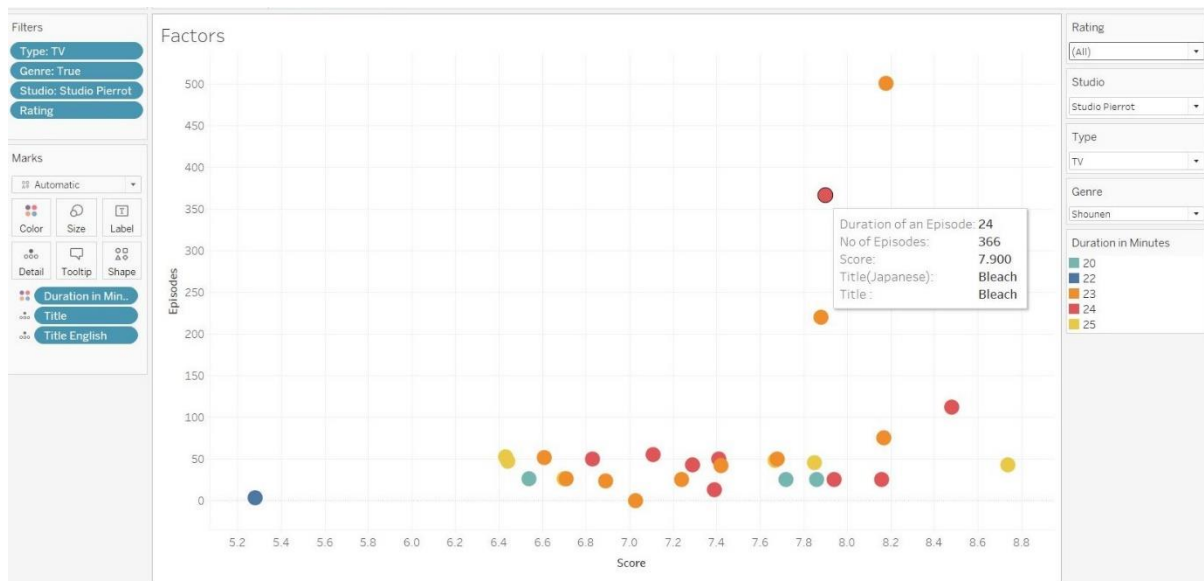


Figure 23: A Scatter plot displaying the factors influencing an Anime's score

Conclusion

Anime has definitely grown to a point where its influence is significant. Although there may be variation in preference regarding the genre we can all agree this is a testament to its range.

There are a few factors like the number of episodes, studio, rating and type have some sort of influence on its success. We found that the episode duration is not a factor as the duration is the same in most cases.

I can say for sure that it did answer to the questions I had. It has also made think about diving deeper into this in the future. When I get hands on better datasets.

Reflection

During my exploration I managed to learn many things that will definitely help me in the future as I work in this field. They are as follows.

- The importance of wrangling / cleaning and how difficult it is to master. I had a first hand experience on how dirty real world data is and the challenges that may arise while dealing with it.
- How spatial data can make a visualization a lot more alluring even though it is difficult as well as have a level of complexity.

In hindsight, I would have loved to convert the state names to its country name while I was wrangling but unfortunately due to the lack of resources I wasn't able to pull it off with the `choroplethAdmin1` package in R. It would have definitely given me more data to analyse. Also I wanted to display the image of the anime in the tooltip. Unluckily, the poster link that was in the dataset itself was invalid. So it wasn't possible for me to do it.

Bibliography

[1] Henerson, Evan, "The Influence of Anime", Keyframe Magazine.

<https://keyframemagazine.org/2020/08/14/the-influence-of-anime/> (Accessed 2020).

[2] Račinský, Matěj, "MyAnimeList Dataset." Kaggle, 2018, doi: 10.34740/KAGGLE/DSV/45582. URL:

<https://www.kaggle.com/azathoth42/myanimelist>.

[3] B. Arel, Vincent, E. Nils, C.J. Yetman, "countrycode: An R package to convert country names and country codes", Journal of Open Source Software, vol. 3, no. 28 ,pp. 848, 2018.

Appendix

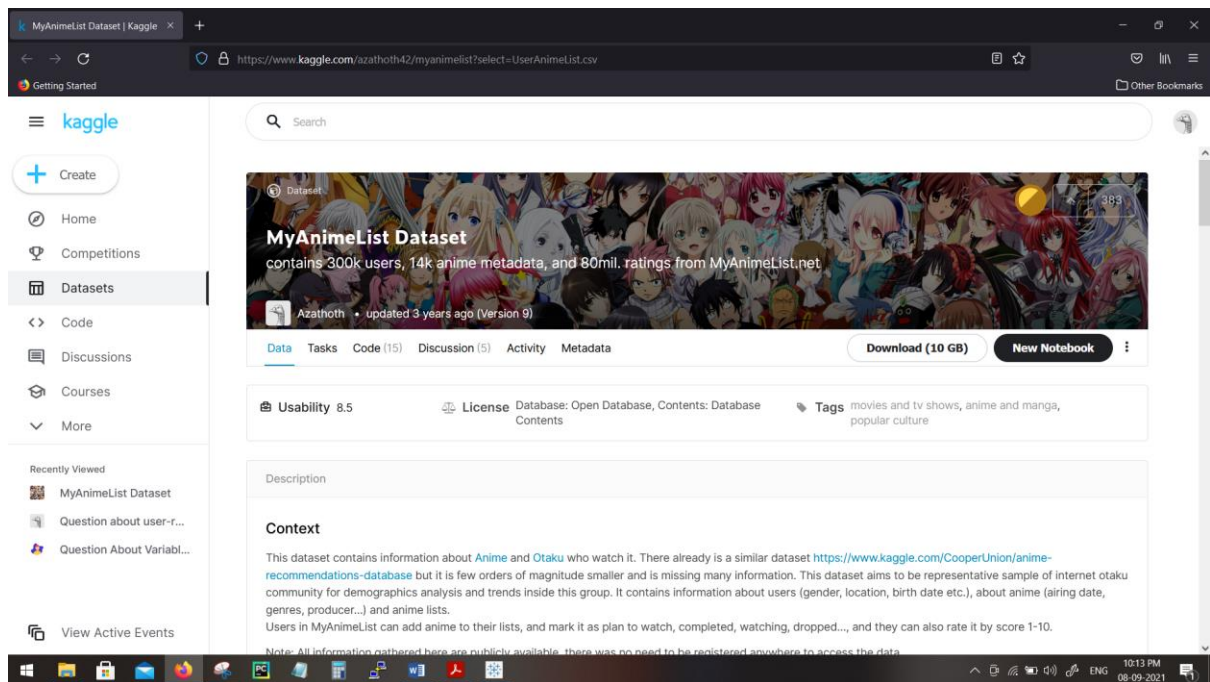


Figure 24: The Kaggle Datasets website screenshot

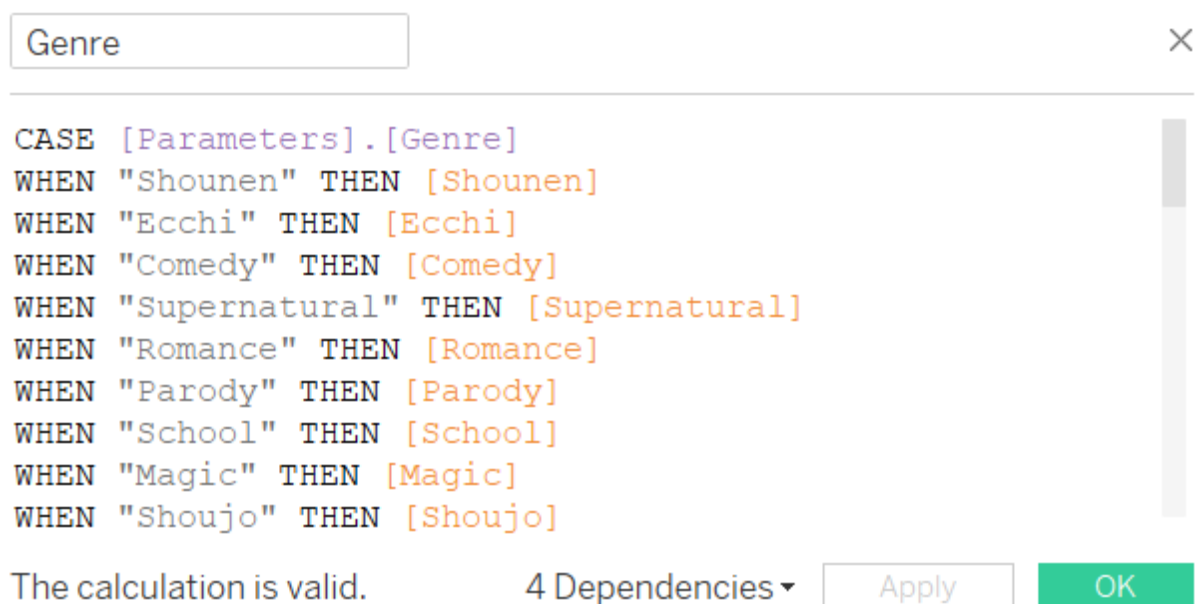


Figure 25: A screenshot of the tableau coding I used for the filters



countrycode

countrycode standardizes country names, converts them into ~40 different coding schemes, and assigns region descriptors. Scroll down for more details or visit the [countrycode CRAN page](#)

If you use countrycode in your research, we would be very grateful if you could cite our paper:

Arel-Bundock, Vincent, Nils Enevoldsen, and CJ Yetman, (2018). countrycode: An R package to convert country names and country codes. Journal of Open Source Software, 3(28), 848, <https://doi.org/10.21105/joss.00848>



Links

Download from CRAN at <https://cloud.r-project.org/package=countrycode>

Browse source code at <https://github.com/vincentarelbundock/countrycode/>

Report a bug at <https://github.com/vincentarelbundock/countrycode/issues>

License

GPL-3

Citation

[Citing countrycode](#)

Developers

Vincent Arel-Bundock

Author, maintainer

[All authors...](#)

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 - [Vectors and data.frames](#)
 - [Flags](#)
 - [Country names in 600+ different languages and formats](#)

Figure 26: A look at countrycode's official website

Source
Light novel ▼
Studio
Madhouse ▼
Rating
PG-13 - Teens 13 or older ▼
Type
TV ▼
Gender
Male ▼
CNT(Title)
<div><div></div></div> <div>1182</div>
Genre
Shounen ▼

Figure 27: A closer look at the filters