

# Report Movie Search Engine

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## Present to

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#### Introduction

According to our group members who like watching movies, our group decided to do 'Movie Search Engine' which is a web search engine for searching movies.

This report explains the details of movie search engines including the problem of the movie search engine, existing relevant system, discussion, and conclusion.

#### **Problems**

#### 1. Dataset

Our group tried to find data from Kaggle that contain all information we needed including:

- Movie's title
- Movies' plot
- Movies' genre
- Original language of movies
- Actors, director
- Movies' rating
- Release date
- Run time of movies
- IMDb id
- Link of movies' poster
- Link to movies's IMDb homepage

But we cannot find all data in one dataset. So, our group decided to use four datasets (four resources) and merge them together into one dataset. After merging all datasets, we have around 1,500 data. You can see all the dataset that our group used in the references.

#### 2. Elastic

The problem we found when implementing Elasticsearch is we cannot build the inverted index for our dataset. Elastics trigger the "File structure cannot be determined" error as shown in the below picture.

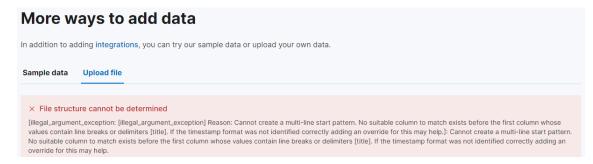


Figure 1 - Elastic error message

We find the issue following the instruction Elastic told us "if the time timestamp format was not identified correctly adding an override for this may help". Since Elastic does not provide a time format that matches the dataset, we try to override the format of the time field with our custom format (dd/MM/yyyy). Finally, the override setting works.

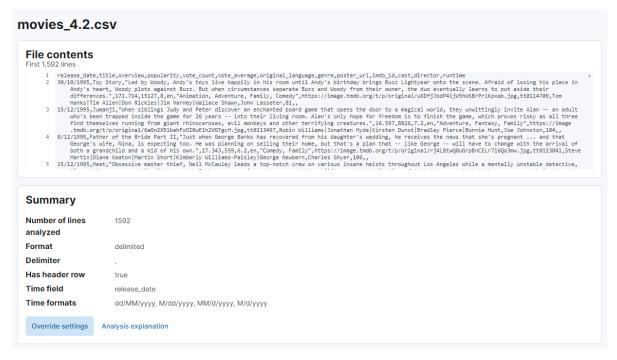


Figure 2 - Elastic with override setting (Time formats)

#### **Existing relevant systems**

Nowadays, there are many movie search engines because people like to search movies to select movies that they are interested in to find more information about the movie. The most popular movie search engines are IMDb and RottenTomatoes.

#### For example:

- https://www.rottentomatoes.com/m/search\_engines
- https://www.imdb.com/search/
- https://www.allmovie.com/advanced-search
- https://www.justwatch.com/uk
- https://reelgood.com/
- https://www.whatismymovie.com/

## Data collection, example documents, and data statistics

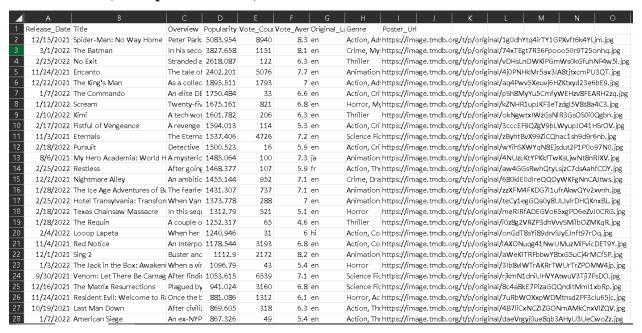


Figure 3 - Dataset [1]

	Α	В	С	D	E E	F	G	н	1	J	K	L	М	N	0	P	Q	R	s	Т	U
1 id		imdb_id	popularity	budget	revenue	original_ti	1 cast	homepage	director	tagline	keywords	overview	runtime	genres	producti	or release_d	vote_cour	vote_aver:	elease_ye	budget_ad	revenue_ad
2	135397	tt0369610	32.98576	1.5E+08	1.51E+09	Jurassic W	Chris Pratt	http://ww	Colin Trev	The park	is monster   c	Twenty-tw	124	Action A	Ad Universa	6/9/2015	5562	6.5	2015	1.38E+08	1.39E+09
3	76341	tt1392190	28.41994	1.5E+08	3.78E+08	Mad Max	Tom Hardy	http://ww	George M	i What a L	o future cha	An apocaly	120	Action	Ad Village R	> #######	6185	7.1	2015	1.38E+08	3.48E+08
4	262500	tt2908446	13.11251	1.1E+08	2.95E+08	Insurgent	Shailene W	http://ww	Robert Scl	One Choi	c based on r	Beatrice P	119	Adventu	re Summit B	n #######	2480	6.3	2015	1.01E+08	2.72E+08
5	140607	tt2488496	11.1731	2E+08	2.07E+09	Star Wars	: Harrison F	http://ww	J.J. Abram	: Every ger	e android   sp	Thirty year	136	Action A	Ad Lucasfiln	1 ########	5292	7.5	2015	1.84E+08	1.9E+09
6	168259	tt2820852	9.335014	1.9E+08	1.51E+09	Furious 7	Vin Diesel	http://ww	James <b>W</b> a	Vengean	ecar race s	Deckard Sh	137	Action 0	Cri Universa	1 4/1/2015	2947	7.3	2015	1.75E+08	1.39E+09
7	281957	tt1663202	9.1107	1.35E+08	5.33E+08	The Rever	Leonardo	http://ww	Alejandro	(n. One w	r father-son	In the 182	156	Western	[ Regency	E #######	3929	7.2	2015	1.24E+08	4.9E+08
8	87101	tt1340138	8.654359	1.55E+08	4.41E+08	Terminato	Arnold Sch	http://ww	Alan Taylo	Reset the	Isaving the	The year is	125	Science	Fic Paramou	n #######	2598	5.8	2015	1.43E+08	4.05E+08
9	286217	tt3659388	7.6674	1.08E+08	5.95E+08	The Marti	Matt Dam	http://ww	Ridley Sco	Bring Him	based on r	During a m	141	Drama /	Ad Twentiet	h #######	4572	7.6	2015	99359956	5.48E+08
10	211672	tt2293640	7.404165	74000000	1.16E+09	Minions	Sandra Bul	http://ww	Kyle Balda	Before G	rı assistant a	Minions St	91	Family /	Ani Universa		2893	6.5	2015	68079970	1.06E+09
11											dream car			Comedy	A Walt Dis	ne 6/9/2015	3935	8	2015	1.61E+08	7.85E+08
12											spy based							6.2			8.1E+08
13											o jupiter spa					2/4/2015		5.2	2015	1.62E+08	1.69E+08
14	264660	tt0470752	6.118847	15000000	36869414	Ex Machin	Domhnall	http://exm	Alex Garla	There is r	nc dancing   a	Caleb, a 26	108	Brama   9	Sci DNA Film	s #######	2854	7.6	2015	13799994	33919846
15		tt2120120									. video gam							5.8			2.24E+08
16											e marvel cor					tu #######		7.4		2.58E+08	
17		tt3460252									o bounty hui					e: ########		7.4			1.43E+08
18											errevenge n					h 1/1/2015		6.1		44159981	3E+08
19											o marvel cor					tu ########		7	2015		4.77E+08
20										-	is cinderella					ne ########		6.8			4.99E+08
21											vi revolution					et unununun		6.5			5.98E+08
22											inventor a					ne ########		6.2			1.92E+08
23							Jake Gyller					Billy "The				rt ########		7.3		27599988	
24											p california					( ########		6.1			4.33E+08
25											u based on r					at ########		5.3			5.24E+08
26						-					rı bank frau					n #######		7.3			1.23E+08
27											e spylseque			Action		n ########		7.1			6.28E+08
28	214756	tt2637276	4.564549	68000000	2.16E+08	Ted 2	Mark Wah	berg   Seth	Seth Maci	Ted is Co	n sperm ban	Newlywed	115	Comedy	Universa		1666	6.3	2015	62559972	1.99E+08

Figure 4 - Dataset [2]

1 A	В	С	D	Ε	F	G	Н		J	K	L	М	N I		P
Title	Year	Summary	Short Sumi	IMDB ID	Runtime	YouTube T	Rating	Movie Pos	Director	Writers	Cast				
Patton Oswalt: Annihilation	2017	Patton Os	Patton Os	tt <b>7</b> 026230	66	4hZi5QaM	7.	1 https://hy	Bobcat G	io Patton Os	Patton Os	walt			
New York Doll	2005	A recoveri	A recoveri	tt0436629	75	jw D04Nsn	7.	https://hy	Greg Whi	it Arthur Kar	Sylvain Sy	lvain			
Mickey's Magical Christmas:	2001	After even	Mickey an	tt0300195	65	uCKwHHft	6.	https://hy	Tony Cra	ig Thomas H	Carlos Ala	zraqui   W	ayne Allwine		
Mickey's House of Villains	2001	The villain:	The villains	tt03293 <b>74</b>	C	JA03ciYt-E	6.	https://hy	Jamie Mi	tc Thomas H	Tony Anse	elmo <b>  W</b> ay	ne Allwine		
And Then I Go	2017	In the crue	In the crue	tt2018111	99	8CdliD6-iF	7.	https://hy	Vincent 6	Gr Brett Hale	Arman Da	rbo Sawy	er Barth		
An Extremely Goofy Movie	2000	It's a big ti	Max goes	tt0208185	79	H8oSvIdA0	6	4 https://hy	Douglas I	M Scott Sper	Bill Farme	r Jason M	1arsden Jeff B	3ennett	
Peter Rabbit	2018	Based on t	Feature ac	tt5117670	95	7Pa_Weid	6.	https://hy	Will Gluc	k Rob Liebe	r Fayssal Ba	azzi James	s Corden		
Love Songs	2007	Julie's boy	A musical i	tt0996605	91	s54vpKAFr	7.	https://hy	Christoph	ne Christoph	Clotilde H	esme Lou	is Garrel   Lud	ivine Sagni	er
89	2017	89 tells the	The true st	tt7614404	91	5hfAExhH	8.	https://hy	Dave Ste	w Lee Dixon	Ian Wrigh	t			
The Foster Boy	2011	The illegiti	A Swiss-G∈	tt2057931	107	E9Qv_XVJ	- 7	1 https://hy	Markus II	m Plinio Bac	Katja Rier	nann   Max	imilian Simor	ischek Ste	efan Ku
Forever My Girl	2018	After being	After being	tt4103724	108	3vqcMr1q	. 6.	4 https://hy	Bethany	A: Bethany A	Abby Ryde	er Fortson	Alex Roe Jes	ssica Roth	9
Tom Segura: Disgraceful	2018	Comedian	Comedian	tt7379330	C	kYYINJM3	7.	https://hy	Jay Karas	Tom Segu	r Tom Segu	ra			
The Secret Rules of Modern	2015	Professor	Professor	tt5818010	58	hpyxmT6K	7.	https://hy	David Bri	ge Marcus du	Sautoy				
Secrets in the Fall	2015	Nothing is	Nothing is	tt2522908	C	eMVw1pH	7.	https://hy	Brittany	G Brittany G	Hollie Sha	y Kyle Ku	oecky Vincen	t Seidle	
Silent Night	2017	Adam une:	Adam une:	tt7133554	100	cA6BUYVk	7.	https://hy	Piotr Dor	n: Piotr Dom	Agnieszka	Suchora	Dawid Ogrodi	nik Tomas	z Ziete
Suicide Squad: Hell to Pay	2018	Task Force	Task Force	tt7167602	86	EPZZvk-wk	7.	https://hy	Sam Liu	Alan Burn	Christian :	Slater <b>  V</b> ar	nessa Williams	S	
Wildling	2018	Anna spen	A blossom	tt5085924	92	eyl1Wf90	6.	https://hy	Fritz Bö	h Fritz Böh	Bel Powle	y Brad Do	ourif Liv Tyler		
The Humanity Bureau	2017	A dystopia	A dystopia	tt6143568	95	kUH8JGhF	6.	https://hy	Rob W. K	ir Dave Schu	Jakob Dav	ies Nicol	as Cage   Sarah	n Lind	
Farewell Ferris Wheel	2016	Farewell, F	Farewell F	tt2349677	72	#NAME?	7	4 https://hy	Miguel 'N	Л. Miguel 'M	.i.G.' Marti	nez			
Don't Talk to Irene	2017	Irene must	Irene must	tt5030452	90	IK48UxWL	6	4 https://hy	Pat Mills	Pat Mills	Anastasia	Phillips B	ruce Gray Mi	chelle McI	.eod
Blood Road	2017	The myste	The myste	tt6900092	96	XCzVqiN9	6.	https://hy	Nicholas	S Mark And	Huyen Ng	uyen lan i	Adamson   Reb	oecca Ruso	h
Andre the Giant	2018	A look at t	A look at t	tt6543420	85	f_jTeuajas	8.	https://hy	Jason He	hi Robin Wri	Cary Elwe	s			
Dead on Arrival	2017	A pharmac	A pharmac	tt5903088	97	8jJ-kC-rq0	( 6.	https://hy	Stephen	C. Stephen C	. Christa B.	Allen D.B	. Sweeney So	ottie Thor	npson
Big Time	2017	Bjarke Inge	BIG TIME f	tt7630164	93	VK0mGdN	1 6.	https://hy	Kaspar A	st Bjarke Ing	Elisabet Ir	ngels			
Adventures in Babysitting	2016	Two teen	Two teen r	tt <b>44</b> 56 <b>8</b> 50	105	BbTTXUM	6.	https://hy	John Schi	ul Tiffany Pa	Nikki Hah	n Sabrina	Carpenter   Sc	ofia Carsor	1
Banana in a Nutshell	2005	An intimat	An intimat	tt1537196	C	6vgD78Z6i	L 6.	https://hy	Roseann	e Roseanne	Liang				
Hostiles	2017	In 1892. af	In 1892. a	tt5478478	134	1M5ci4Un	r 7.	https://hv	Scott Co	or Scott Coo	Ava Coop	er l Rosam	und Pike   Sco	tt Shepher	d

Figure 5 - Dataset [3]

	А	В	c	D	E	F	G	н	1	J	к	ι	М	N	О
1	Release_Date	Movies' Title	Movies' Plot	Movies' Popularity	Vote_Count	Vote_Average	Language	Genre	Poster_Url	match_imdb_id	match_cast	match_director	match_runtime	merge_summary	
2	5/13/2015	Mad Max: Fu	r An apocalyptic story set in the	89.139	19090	7.5	en	Action,	https://ima	tt1392190	Tom Hardy	George Miller	120	In a post-apocalyp	tic waste
3	3/18/2015	Insurgent	Beatrice Prior must confront	78.099	8780	6.4	en	Action,	https://ima	tt2908446	Shailene Wo	Robert Schwentk	119	Beatrice Prior mus	st confro
4	12/15/2015	Star Wars: Th	Thirty years after defeating th	77.811	16772	7.3	en	Action,	https://ima	tt2488496	Harrison For	dJJ. Abrams	136	Three decades after	erthe En
5	4/1/2015	Furious 7	Deckard Shaw seeks revenge	105.485	8973	7.3	en	Action,	https://ima	tt2820852	Vin Diesel Pa	James Wan	137	Deckard Shaw see	ks revenį
6	12/25/2015	The Revenant	In the 1820s, a frontiersman,	53.111	15554	7.5	en	Wester	r https://ima	tt1663202	Leonardo Dio	Alejandro GonzÃj	156	A frontiersman on	a fur tra
7	6/23/2015	Terminator G	The year is 2029. John Conno	108.488	7089	5.9	en	Science	https://ima	a,tt1340138	Arnold Schwa	a Alan Taylor	125	When John Conno	r, leader
8	9/30/2015	The Martian	During a manned mission to I	68.531	16724	7.7	en	Drama,	https://ima	tt3659388	Matt Damon	Ridley Scott	141	An astronaut beco	mes stra
9	6/17/2015	Minions	Minions Stuart, Kevin and Bo	45.208	9143	6.4	en	Family,	https://ima	tt2293640	Sandra Bullo	Kyle Balda   Pierre	91	Minions Stuart, Ke	vin, and
10	6/9/2015	Inside Out	Growing up can be a bumpy i	108.817	17586	7.9	en	Animat	https://ima	tt2096673	Amy Poehler	Pete Docter	94	After young Riley i	s uproot
11	10/26/2015	Spectre	A cryptic message from Bond	58.265	8944	6.5	en	Action,	https://ima	a, tt2379713	Daniel Craig	(Sam Mendes	148	A cryptic message	from 00
12	2/4/2015	Jupiter Ascen	In a universe where human g	64.915	5811	5.4	en	Science	https://ima	tt1617661	Mila Kunis C	l Lana Wachowski	124	A young woman d	iscovers
13	12/16/2014	Ex Machina	Caleb, a coder at the world's	27.676	11034	7.6	en	Drama,	https://ima	tt0470752	Domhnall Gle	Alex Garland	108	A young programs	mer is sel
14	7/16/2015	Pixels	Video game experts are recru	79.688	6366	5.7	en	Action,	https://ima	a,tt2120120	Adam Sandle	r Chris Columbus	105	When aliens misin	terpret v
15	4/22/2015	Avengers: Ag	When Tony Stark tries to jum	163.611	19473	7.3	en	Action,	https://ima	tt2395427	Robert Down	Joss Whedon	141	When Tony Stark a	and Bruc
16	12/25/2015	The Hateful E	i Bounty hunters seek shelter f	32.716	11871	7.7	en	Drama,	https://ima	tt3460252	Samuel L. Jac	Quentin Tarantin	167	In the dead of a W	yoming '
17	12/16/2014	Taken 3	Ex-government operative Bry	119.377	4792	6.2	en	Thriller,	https://ima	a,tt2446042	Liam Neeson	Olivier Megaton	109	Ex-government op	erative E
18	7/14/2015	Ant-Man	Armed with the astonishing a	85.255	16802	7.1	en	Science	https://ima	tt0478970	Paul Rudd   N	1 Peyton Reed	115	Armed with a supe	ar-suit wi
19	11/18/2015	The Hunger G	With the nation of Panem in a	75.942	10412	6.9	en	Action,	https://ima	tt1951266	Jennifer Law	Francis Lawrence	136	As the war of Pane	am escala
20	5/19/2015	Tomorrowlar	Bound by a shared destiny, a	47.761	6056	6.2	en	Advent	https://ima	tt1964418	Britt Roberts	c Brad Bird	130	Bound by a shared	destiny,
21	3/24/2015	Southpaw	Billy "The Great" Hope, the re	101.201	4644	7.4	en	Action,	( https://ima	tt1798684	Jake Gyllenha	Antoine Fugua	123	A champion boxer	fightstc
22	5/27/2015	San Andreas	In the aftermath of a massive	101.043	7077	6.2	en	Action,	https://ima	tt2126355	Dwayne John	Brad Peyton	114	In the aftermath o	f a mass
23	2/11/2015	Fifty Shades o	When college senior Anastasi	150.025	10087	5.9	en	Drama,	https://ima	tt2322441	Dakota Johns	Sam Taylor-Johns	125	Literature student	Anastas
24	12/11/2015	The Big Short	The men who made millions f	29.765	7289	7.3	en	Comed	https://ima	tt1596363	Christian Bale	Adam McKay	130	In 2006-7 a group	of invest
25	7/23/2015	Mission: Impo	Ethan and team take on their	48.086	7178	7.2	en	Action,	https://ima	tt2381249	Tom Cruise	Christopher McQ	131	Ethan and team ta	ke on the
26	c/sc/ss4	T 15		430.03				~ 1	111 22			edar ed	***	a	- · ·

Figure 6 - Dataset [4]

From figure 1, 2, and 3 there are three dataset that our group found on Kaggle. Our group merged three dataset to one dataset which is figure 4. In figure 4, it has around 1,500 data in the dataset.

## **Tools and software**

## • Excel (data)

Our group uses Excel for collecting data and use in the Elasticsearch for calling data when the search engine is working.

## • Visual Studio Code (HTML and CSS)

Our group uses HTML and CSS for the interface of our movie search engine. HTML is the standard markup language for Web pages [4]. CSS is used to format the layout of a webpage which our group uses both inner CSS and external CSS [5].

## System diagram

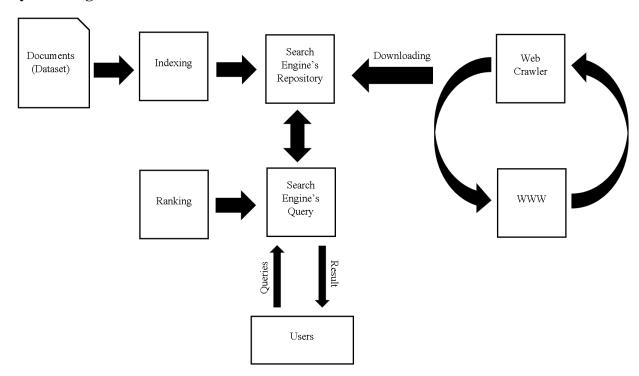


Figure 7 Movie Search System Diagram

## **Snapshots of the system**

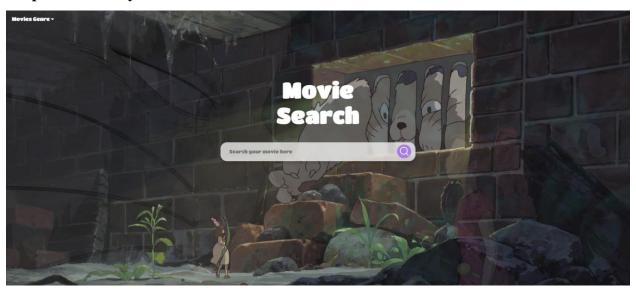


Figure 8 - Movie search page



Figure 9 - Search result page (Toy Story)

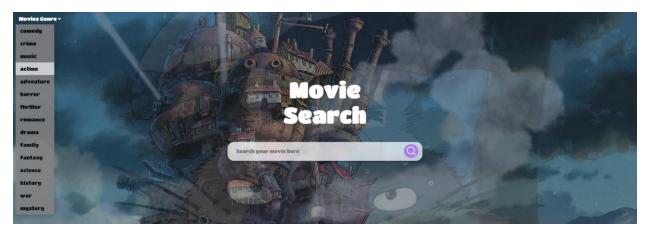


Figure 10 - Movie category selection

## **Example step-by-step search sessions**

## 1. One word query



Figure 11 - Step 1, put a keyword (sense) in the search box and press enter



Figure 12 - Step 2, the results of the keyword (sense) has shown on the result page

## 2. Multiple words query

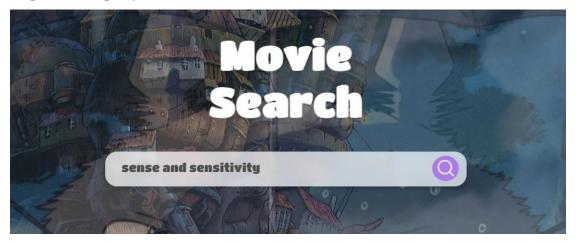


Figure 13 - Step 1, put multiple words (sense and sensitivity) in the search box and press enter

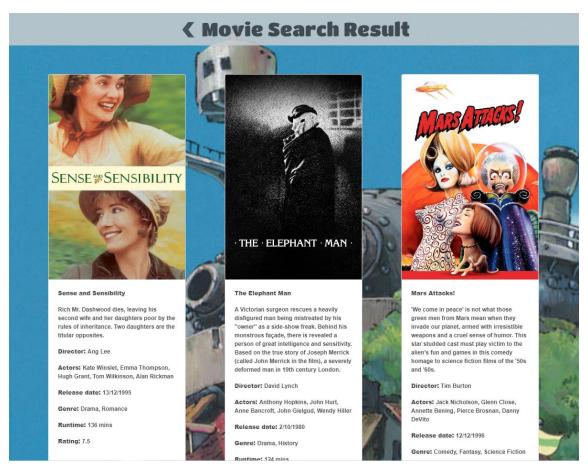


Figure 14 - Step 2, the results of the keyword (sense and sensitivity) has shown on the result page

## 3. Partial match

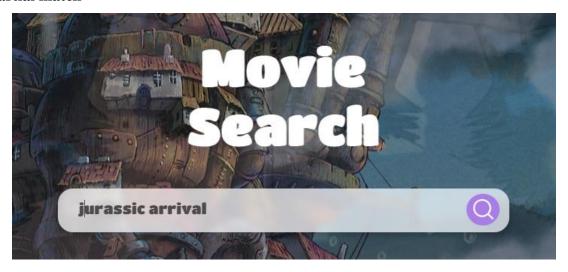


Figure 15 - Step 1, put multiple different words (jurassic arrival) in the search box and press enter

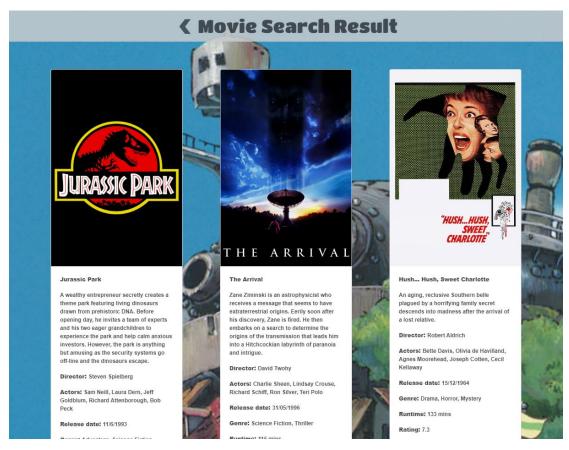


Figure 16 - Step 2, the results of the keyword (jurassic arrival) has shown on the result page

## 4. Ranking

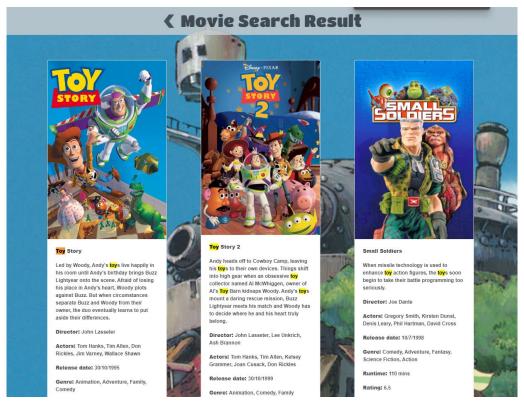
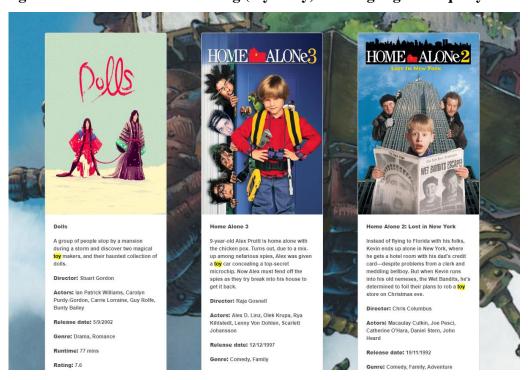


Figure 17 - the results of searching (toy story) with highlight the query word



## Figure 18 - the results of searching (toy story) with highlight the query word (continue) Discussion

## **Limitations of your system**

Since our movie search engine is the very first demo version, the result of the search is not efficient enough. Our search engine only looks up the title, overview, director, and cast fields of the dataset (also the movie's genre but we make it like a filter search, not a query search). The efficiency of the search is not good enough because we did not optimize the search threshold. We think we can improve search performance better than this in a later version. The last thing is the search engine does not provide spelling corrections. the user must type every word carefully.

## Technical difficulties, challenges, and lessons learned

There are technical difficulties that our group met including:

#### Interface

For the interface, it is difficult to create a web page by using HTML and CSS to have the same interface. When we open it on another laptop, it is always different. So, we have to create a web page to look the same as much as we can.

## Server Downloading

For the server downloading, it is really slow to download the image. You have to wait for the image download which takes a while. Also, background images of the website are not continuing from the previous image.

Our group learned many things from creating movie search engines. These are what we learn from create movie search engine:

- Learning how to prepare data by obtaining data from various sources and combining this together before we can process and retrieve our final dataset.
- To ensure that our dataset has high quality and data integrity, we learned how to clean and eliminate duplicate data, transform the data type, correct errors, and deal with missing values.
- To be able to become familiar with new tools like Elastic search, which enables us to quickly store, search, and analyze large amounts of data and return valuable results.
- The management of teams is the most significant lesson we have learned. Our team, which is made up of individuals with diverse areas of expertise, has divided the particular tasks

for each member, including front-end, back-end, and database management. As a result, our work was completed in a short period of time.

## **Opportunities for future improvements**

Our group discussed about what we need to improve on our movie search engine. These are what we need to improve or create more functionalities:

## • Old movies searching

Most movie search engines don't have many old movies below 2000. Some people like to watch old movies. If possible, it might be good for our movie search engine.

## • Movies searching by photo

Some people might have some picture of the movie, but they cannot remember the movie's title. If we can add this feature to our movie search engine, it will be more convenient and a lot easier for finding movie's information.

## Conclusion

In this project, we implemented a simple movie search (search engine) that can search a movie by title, storyline, director, and cast. We built the project with many tools: Microsoft Excel - used to handle records in the dataset, Elastic and Kibana - used to build and manipulate the inverted index via a beautiful configuration interface, and Flask - used to design and create the interface of our web-based search engine. In this project, we ran into many unexpected problems that we had to find a way to solve. For instance, the problem in the dataset preparation is that we cannot find the dataset that contains the information we want about movies, so we have to merge many datasets into the perfect one before we go to create the inverted index. Also the problem of building the inverted index, the dataset we have uploaded to Elastic triggers an error about the field determined, then we try to follow the instructions of its and try to customize the format of the time field after that, an error has gone. In this project, we have learned many things which can help us to do a job in the real world such as work breakdown structure, collaboration in the team, and problem-solving skills. Eventually, we accomplish many things more than just a search engine project. Thank you.

#### References

- [1] Hasan, D.I.R. (2022) *9000+ movies dataset, Kaggle*. Available at: https://www.kaggle.com/datasets/240bbdc8f00ac47a8b212c9815991ba2d1d9a382c8bc6a103b90 6f43e7aba303 (Accessed: November 23, 2022).
- [2] Tanyu, K. (2018) *Movie dataset, Kaggle*. Available at: https://www.kaggle.com/datasets/kayscrapes/movie-dataset (Accessed: November 23, 2022).
- [3] Shakir, J. (2018) *TMDB Movies Dataset*, *Kaggle*. Available at: https://www.kaggle.com/datasets/juzershakir/tmdb-movies-dataset (Accessed: November 23, 2022).
- [4] Hasan, D.I.R. (2022) *9000+ movies dataset, Kaggle*. Available at: https://www.kaggle.com/datasets/240bbdc8f00ac47a8b212c9815991ba2d1d9a382c8bc6a103b90 6f43e7aba303 (Accessed: November 23, 2022).
- [5] *HTML styles CSS* (no date) *HTML Styles CSS*. Available at: https://www.w3schools.com/html/html\_css.asp (Accessed: November 23, 2022).