Identifying Medical Drama Narratives: a Multi-Output Regression approach with CNNs and BERT

Alice Fedotova

- Medical dramas are a popular genre of television programming that depicts the lives and work of medical professionals
- It is one of the most popular and longestrunning genres on global television



- In the field of media studies, content analysis has long been employed as a methodology for the study of audiovisual products. A central aspect of content analysis is coding, which consists in assigning units of analysis to categories for the purpose of describing and quantifying phenomena of interest (Krippendorff, 1995).
- Previous research has identified three fundamental categories or "isotopies" that characterize the medical drama genre: the **medical case**, which refers to the new medical cases that are presented in each episode; the **professional plot**, which focuses on the portrayal of medical professionals and their work; and the **sentimental plot**, which focuses on the personal lives and relationships of the characters (Rocchi and Pescatore, 2019).

Modeling narrative features in TV series: coding and clustering analysis

Marta Rocchi 🗠 & Guglielmo Pescatore



https://doi.org/10.1057/s41599-022-01352-9

- Content analysis requires expert knowledge and trained annotators
- The task of manually identifying *segments* that pertain to the isotopies, combined with the time required to assign the appropriate codes, makes content analysis a significantly time-consuming process (Rocchi and Pescatore, 2022)

- Content analysis requires expert knowledge and trained annotators
- The task of manually identifying *segments* that pertain to the isotopies, combined with the time required to assign the appropriate codes, makes content analysis a significantly time-consuming process (Rocchi and Pescatore, 2022)
- Is it possible to predict the narrative composition of a medical drama by looking only at the subtitles?

	A	В	с	D	E	F	G	н	I	J	к	L	м
113	Season	Codice	N_segmento	Inizio	Fine	Durata	PP	PP_rel	SP	SP_rel	МС	MC_rel	Note
114	GAS13	GAS13E03	1	00:00:00	00:00:01	00:00:01	NA		NA		NA		Contes
115	GAS13	GAS13E03	2	00:00:01	00:01:04	00:01:03	0		6	SP3,SP7	0		Pg+mo
116	GAS13	GAS13E03	3A	00:01:04	00:01:20	00:00:16	0		6	SP3	0		
117	GAS13	GAS13E03	3B	00:01:20	00:01:59	00:00:39	0		6	SP3	0		
118	GAS13	GAS13E03	4A	00:01:59	00:02:26	00:00:27	0		6	SP3	0		
119	GAS13	GAS13E03	4B	00:02:26	00:02:40	00:00:14	6	PP1	0		0		
120	GAS13	GAS13E03	4C	00:02:40	00:03:13	00:00:33	0		6	SP3	0		
121	GAS13	GAS13E03	5	00:03:13	00:04:20	00:01:07	0		6	P1,SP2,SP	0		
122	GAS13	GAS13E03	6	00:04:20	00:05:25	00:01:05	0		0		6	MC1,MC3	3 monta
123	GAS13	GAS13E03	7	00:05:25	00:06:37	00:01:12	0		0		6	MC4,MC7	7
124	GAS13	GAS13E03	8	00:06:37	00:07:14	00:00:37	6	PP1,PP6	0		0		

1563	GAS13	GAS13E24	0:35:44	0:36:05	0:00:21	0		4	PP1	2	SN7
1564	GAS13	GAS13E24	0:36:05	0:36:29	0:00:24	0		0		6	SP7
1565	GAS13	GAS13E24	0:36:29	0:36:33	0:00:04	0		0		6	SP10
1566	GAS13	GAS13E24	0:36:33	0:36:50	0:00:17	0		2	PP1	4	SP1
1567	GAS13	GAS13E24	0:36:50	0:36:55	0:00:05	0		0		6	SP10
1568	GAS13	GAS13E24	0:36:55	0:37:22	0:00:27	5	MC1	1	PP1	0	
1569	GAS13	GAS13E24	0:37:22	0:37:45	0:00:22	4	MC4	2	PP1	0	
1570	GAS13	GAS13E24	0:37:45	0:38:36	0:00:51	0		0		6	SP10
1571	GAS13	GAS13E24	0:38:36	0:39:08	0:00:32	0		6	PP1 PP5	0	
1572	GAS13	GAS13E24	0:39:08	0:39:15	0:00:07	0		6	PP1	0	

1

00:00:04,671 --> 00:00:06,739 Meredith: Every patient who gets bad news

2 00:00:06,773 --> 00:00:08,207 <i>wants to be the exception...</i>

3
00:00:08,241 --> 00:00:10,709
<i>the one in a million, the miracle.</i>

4

00:00:12,016 --> 00:00:14,279 <i>But surgeons don't believe in miracles.</i>

5 00:00:14,314 --> 00:00:16,615 <i>We have to make them happen.</i>

6

00:00:16,649 --> 00:00:17,983 Are you okay?

7

8

00:00:18,018 --> 00:00:19,651 He has all the power now.

00:00:19,686 --> 00:00:21,787 It's wrong. I'm his boss.

	Segment start	Segment end	Subtitle start	Subtitle end	РР	SP	мс	Subtitle text
0	00:00:01	00:01:04	00:00:04.671	00:00:06.739	0	6	0	Meredith: Every patient who gets bad news
1	00:00:01	00:01:04	00:00:06.773	00:00:08.207	0	6	0	wants to be the exception
2	00:00:01	00:01:04	00:00:08.241	00:00:10.709	0	6	0	the one in a million, the miracle.
3	00:00:01	00:01:04	00:00:12.016	00:00:14.279	0	6	0	But surgeons don't believe in miracles.
4	00:00:01	00:01:04	00:00:14.314	00:00:16.615	0	6	0	We have to make them happen.
21143	00:41:23	00:41:43	00:41:23.297	00:41:26.065	0	0	0	Sometimes, we wake up, we face our fears
21144	00:41:43	00:41:47	00:41:42.849	00:41:45.149	0	0	6	We take them by the hand.
21145	00:41:59	00:42:10	00:42:01.299	00:42:03.800	0	0	6	And we stand there, waiting
21146	00:41:59	00:42:10	00:42:03.836	00:42:06.169	0	0	6	hoping, ready
21147	00:41:59	00:42:10	00:42:06.205	00:42:08.705	0	0	6	for anything.

	Instances
Train	55,567
Dev	11,907
Test	11,908
Overall	79,382

	PP	SP	MC	Total
Train	17,391	34,555	25,250	77,196
Dev	3,773	7,343	5,415	16,531
Test	3,701	7,384	5,449	16,534
Overall	24,865	49,282	36,124	110,271

Instances in the dataset divided by plot type: SP is the most prevalent in the case of Grey's Anatomy

Task Definition

- Given the structure of the resulting dataset, the task of predicting the narratives along with the corresponding weights was framed as **multi-output regression**, also known as **multi-target regression**
- It is a combination of multiple regression problems, where each output is predicted using the same input data
- In the case of the present work, the outputs are the weights relative to the three plot types, whereas the input is the text of a subtitle

CNNs

- Embeddings: GloVe
- **Preprocessing:** lowercasing and *normalization*

• Tuning:

 $CNN_1 \rightarrow manual$

 $CNN_2 \rightarrow automated$

• **Output layer:** sigmoid with three outputs

Parameters	Settings	CNN ₁	CNN ₂
Conv1D kernel size	€ [1, 2, 3, 4, 5]	3	2
Conv1D filters	∈ [50 250] with a step size of 25	250	50
Dense units	∈ [32 256] with a step size of 32	250	352
Dropout value	€ [0.05, 0.1, 0.2, 0.3]	0.2	0.1
Intermediate layers	€ [1, 2, 3]	1	2
Loss function	MAE, MSE	MSE	MSE
Adam learning rate	€ [0.002, 0.001, 0.0001]	0.001	0.001

Subtitle	PP	SP	MC
amelia invited riggs to dinner at our house.	0	1	0
it's about the whole healthcare system, not this place.	1	0	0
noelle webb, 43, complains of abdominal pain	0	0	1
i'm the chief of general, i loved working with you,	0.5	0.5	0
if I have to look him in the eye and tell him i blew it	0	0.83	0.16
yes, well, the medical community and i are in a fight.	0.33	0.66	0
why? my patient is terrified.	0.33	0	0.66
yeah, not by you. page surgery.	0.66	0	0.33
whatever. he bends his rules all the time to save his own patients.	0.5	0.5	0

BERT

	Batch size						
Epochs	16	32					
2	0.2832	0.2824					
3	0.2898	0.2868					
4	0.2891	0.2882					

Performance on the validation set with a learning rate of 2e-5, expressed in MAE

- Model: bert-base-uncased
- Preprocessing: BertTokenizer
- **Tuning:** experimented with the parameters suggested by Devlin et al. (2019)
- Parameters: 2 epochs, batch size of 32, learning rate of 2e-5
- Output layer: sigmoid with three outputs

data_type	subtitle_text	mc_pred	sp_pred	pp_pred	mc	sp	рр	
test	wh am i gonna die?	0.5616	0.0093	0.0005	1.0	0.0	0.0	0
test	she'd be dead right now if it weren't for you	0.5083	0.0571	0.0010	0.0	0.0	1.0	1
test	she's my new best friend.	0.0009	0.8240	0.0001	0.0	0.0	1.0	2
test	like the kind we feel when we win.	0.0006	0.9800	0.0003	0.0	0.0	1.0	3
test	just him and me against the world.	0.0025	0.5105	0.0002	0.0	0.0	1.0	4
test	you realize how this looks, don't you?	0.0260	0.3212	0.0002	1.0	0.0	0.0	7827
test	i'm prioritizing my family, my sister.	0.0012	0.6068	0.0002	1.0	0.0	0.0	7828

data_type	subtitle_text	mc_pred	sp_pred	pp_pred	mc	sp	рр	
test	wh am i gonna die?	0.5616	0.0093	0.0005	1.0	0.0	0.0	0
test	she'd be dead right now if it weren't for you	0.5083	0.0571	0.0010	0.0	0.0	1.0	1
test	she's my new best friend.	0.0009	0.8240	0.0001	0.0	0.0	1.0	2
test	like the kind we feel when we win.	0.0006	0.9800	0.0003	0.0	0.0	1.0	3
test	just him and me against the world.	0.0025	0.5105	0.0002	0.0	0.0	1.0	4
test	you realize how this looks, don't you?	0.0260	0.3212	0.0002	1.0	0.0	0.0	7827
test	i'm prioritizing my family, my sister.	0.0012	0.6068	0.0002	1.0	0.0	0.0	7828

Results and Evaluation

		R	² ↑		MAE 🗸				RMSE ↓			
	PP	SP	MC	All	PP	SP	MC	All	PP	SP	MC	All
Baseline	0.00	0.00	0.00	0.00	0.26	0.39	0.37	0.34	0.32	0.43	0.41	0.15
CNN 1	-0.05	-0.05	-0.06	-0.05	0.24	0.38	0.35	0.32	0.33	0.44	0.42	0.16
CNN 2	0.02	0.05	0.03	0.03	0.26	0.37	0.36	0.33	0.32	0.41	0.41	0.15
BERT	0.05	0.11	0.07	0.08	0.19	0.35	0.31	0.28	0.37	0.47	0.46	0.19

Conclusion and Future Work

- The problem of predicting the narrative composition of Grey's Anatomy was addressed with three multioutput regression models in order to learn both the plot types and the weights, at the level of a subtitle
- The best performing model, based on BERT, obtained an R² score of 0.08; CNN₂ performed better than CNN₁
- While visual and audio information have been widely studied, textual clues are less explored (Weng at al., 2021)
- As demonstrated by Li et al. (2021), leveraging both the video and the subtitles achieves competitive results on video understanding tasks