Sentiment Analysis of Movie Reviews Using Multiple Machine Learning Methods

Jack St. Clair & Zachary Toner



1/10

The Worst Film Ever Made

MattClayton92 27 April 2016

When someone calls a film "terrible", they're usually either exaggerating or have no concept of what a terrible film truly is. The Room embodies that concept. It's not just poorly written, poorly acted, poorly directed - poorly everything - it honestly looks like



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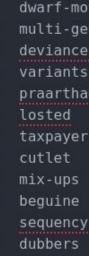
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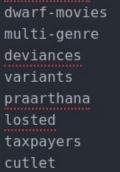
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4 0:12 1:4 2:6 3:4 4:5 5:4 6:2 7:2 8:3 9:2 10:3 11:4 12:1 13:3 14:2 16:3 17:2 18:1 22:1 23:1 24:1 25:1 29:1 30:1 31:1 32:1 33:1 38:2 41:1 43:2 45:1 50:1 53:1 1505:1 1522:1 1590:1 1735:1 1793:1 1920:1 2081:1 2106:1 2268:2 2631:1 3307:1 3308:1 3316:1 3663:1 4143:1 4172:1 4643:1 5458:1 5703:2 6089:1 6729:1 7544:1 824 10 1:1 2:2 3:2 4:1 5:2 6:1 7:2 9:1 10:1 12:1 13:1 14:1 17:1 22:1 23:1 114:1 115:1 132:1 148:2 281:1 290:1 306:1 327:1 543:1 565:1 594:1 667:1 677:1 681:1 687 8 0:16 1:11 2:12 3:14 4:11 5:6 6:4 7:9 8:14 9:2 10:1 11:2 13:4 14:2 16:4 17:13 18:4 19:2 20:3 21:5 22:3 23:3 25:2 26:1 27:11 28:4 32:3 34:3 35:1 36:5 37:2 39 313:1 341:1 345:1 357:1 363:1 369:1 384:1 393:1 427:2 435:2 439:1 461:1 479:1 483:1 489:2 511:1 548:1 570:2 579:1 732:1 796:1 838:1 845:1 868:1 942:1 1031:2

6774:2 7908:1 8095:1 8105:1 8938:1 10013:1 10067:1 10238:1 11469:1 11544:2 13079:1 15181:1 16140:1 84632:1





K-Nearest Neighbors

Binary	Predicted Bad Reviews	Predicted Good Reviews
True Bad Reviews	550	471
True Good Reviews	312	667

Accuracy: 0.6085

Hyperparameters
Weights: Uniform
Nearest neighbors: 5

K-Nearest Neighbors

Accuracy: 0.2465

Hyperparameters
Weights: Uniform
Nearest neighbors: 5

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True Labels

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Predicted	Labels							
Multiclass	1	2	3	4	7	8	9	10
1	211.	25.	7.	20.	6.	39.	12.	99.
2	90.	8.	2.	9.	6.	25.	9.	47.
3	63.	14.	6.	20.	7.	26.	5.	62.
4	60.	4.	9.	22.	9.	36.	10.	53.
7	51.	8.	4.	13.	12.	37.	4.	39.
8	50.	5.	8.	24.	5.	54.	12.	58.
9	30.	8.	7.	15.	12.	40.	10.	69.
10	109.	10.	7.	29.	5.	56.	18.	170.

Random Forests

Binary	Predicted Bad Reviews	Predicted Good Reviews
True Bad Reviews	847	174
True Good Reviews	179	800

Accuracy: 0.8235

Hyperparameters
Max_Features: sqrt

Random Forests

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Accuracy: 0.363

Hyperparameters
Max_Features: sqrt

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Predicted Lahele

Predicted	Labels							
Multiclass	1	2	3	4	7	8	9	10
1	329.	3.	3.	5.	2.	6.	Θ.	71.
2	133.	2.	6.	3.	3.	4.	Θ.	45.
3	124.	Θ.	3.	6.	7.	8.	Θ.	55.
4	79.	3.	5.	15.	9.	14.	Θ.	78.
7	40	1.	2.	9	6.	33.	0.	77.

32. 1. 1. 3. 10. 42. 2. 125.

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True Labels

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Support Vector Machines

Binary	Predicted Bad Reviews	Predicted Good Reviews
True Bad Reviews	811	210
True Good Reviews	180	799

Accuracy: 0.805

Hyperparameters

C: 1.0

Gamma: 1/n Features

Hyperparameters Support Vector Machines Accuracy: 0.3485 C: 1.0 Gamma: 1/n features **Predicted Labels** Multiclass 9 10 23. 20. 30. 14. 5. 7. 20. 24. 27. 7. 5. 5. 16. 21. 22. 37. 13. 12. 4. 3 13. 26. 52. 10. 26. 10. 16. 24. 26. 30. 17. 36. 11. 29. 27. 33. 8 Labels 1. 12. 21. 17. 29. 22. 9 16. 12. 35. 42. 232. 10

