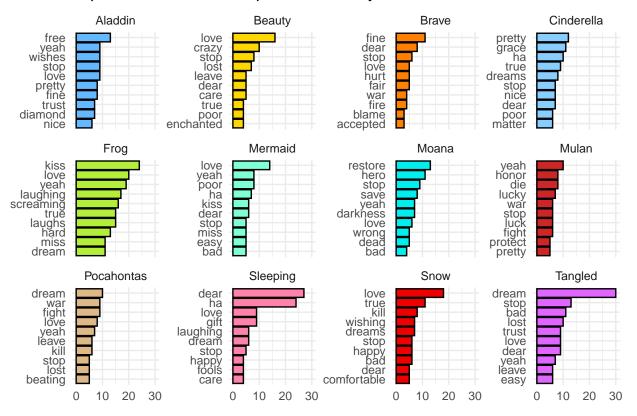
## A quick and easy way to show off the fun stuff

```
# read in subtitle files and create a `movie` column for identification
animated <- rbindlist(list(</pre>
  data.table(read_srt(path = "MovieSubtitles/Animated/Aladdin.srt", collapse = " "))[, movie := "Aladdin.srt", collapse = " "))
  data.table(read_srt(path = "MovieSubtitles/Animated/Beauty and the Beast.srt", collapse = " "))[, mov
  data.table(read_srt(path = "MovieSubtitles/Animated/Brave.srt", collapse = " "))[, movie := "Brave"],
  data.table(read_srt(path = "MovieSubtitles/Animated/Cinderella.srt", collapse = " "))[, movie := "Cinderella.srt",
  data.table(read_srt(path = "MovieSubtitles/Animated/The Princess and the Frog.srt", collapse = " "))[
  data.table(read_srt(path = "MovieSubtitles/Animated/The Little Mermaid.srt", collapse = " "))[, movie
  data.table(read_srt(path = "MovieSubtitles/Animated/Moana.srt", collapse = " "))[, movie := "Moana"],
  data.table(read_srt(path = "MovieSubtitles/Animated/Mulan.srt", collapse = " "))[, movie := "Mulan"],
  data.table(read_srt(path = "MovieSubtitles/Animated/Pocahontas.srt", collapse = " "))[, movie := "Poc
  data.table(read_srt(path = "MovieSubtitles/Animated/Sleeping Beauty.srt", collapse = " "))[, movie :=
  data.table(read_srt(path = "MovieSubtitles/Animated/Snow White and The Seven Dwarfs.srt", collapse =
  data.table(read_srt(path = "MovieSubtitles/Animated/Tangled.srt", collapse = " "))[, movie := "Tangle
animated[sample(1:nrow(animated), size = 6)]
##
              start
                                                        subtitle
                                                                      movie
      756 4150.104 4153.148
                                                 -No! -Come on !
                                                                      Mulan
## 2: 594 2933.724 2936.143 What's gotten into you, fella? Oh.
                                                                    Mermaid
## 3: 1218 3899.354 3901.396
                                   For you, it's gonna be tough
                                                                       Frog
                                                      Thank you. Cinderella
## 4: 775 3263.601 3264.521
## 5: 1075 4589.520 4591.238
                                               You can't fix it!
                                                                      Moana
## 6: 372 1966.048 1968.259 That's why the wisps led me here.
                                                                      Brave
# generates longer data using each of the three dictionaries
# each word is on its own row
aniAfinnWordLonger <- animated %>%
  unnest_tokens(word, subtitle) %>%
  .[afinn, on = "word", nomatch = 0]
aniAfinnWordLonger[sample(1:nrow(aniAfinnWordLonger), size = 6)]
##
             start
                                 movie startScale endScale song
                                                                       word value
                        end
## 1: 748 3154.393 3155.121 Cinderella 0.7403409 0.7393216 FALSE
                                                                     please
## 2: 541 3556.015 3559.477
                                  Snow 0.7036807 0.7035562 FALSE goodness
                                                                                3
## 3: 449 1503.585 1506.129
                                  Frog 0.2602189 0.2599959 FALSE
                                                                       poor
                                                                                -2
## 4: 723 3838.752 3841.962
                                 Mulan 0.7818999 0.7821767 FALSE
                                                                       yeah
                                                                                1
## 5: 741 3142.387 3145.936
                               Aladdin 0.6359557 0.6361464 FALSE
                                                                      laugh
                                                                                1
## 6: 284 1567.858 1570.569
                               Mermaid 0.3235100 0.3233132 FALSE pressure
                                                                                -1
aniBingWordLonger <- animated %>%
  unnest_tokens(word, subtitle) %>%
  .[bing, on = "word", nomatch = 0]
aniBingWordLonger[sample(1:nrow(aniBingWordLonger), size = 6)]
```

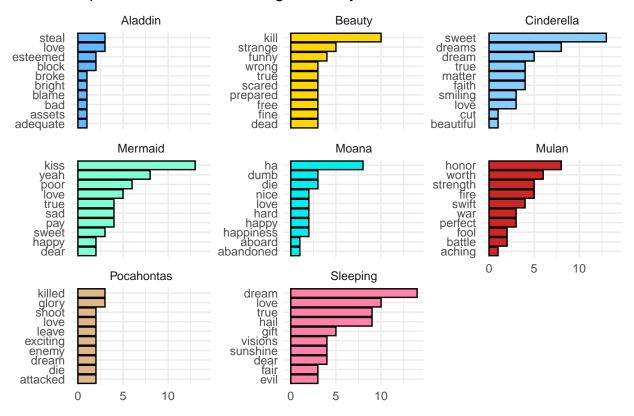
```
movie startScale endScale song
                        end
                                                                      word
## 1: 675 3314.021 3318.108 Mermaid 0.70217454 0.7020031 FALSE
                                                                      easy
## 2: 356 1846.910 1850.475 Tangled 0.33054471 0.3305995 FALSE triumphant
## 3: 1026 5434.158 5435.857 Tangled 0.99191696 0.9916546 FALSE
## 4: 401 2069.026 2073.656
                              Brave 0.36142879 0.3617203 FALSE attractive
       30 127.547 131.096 Aladdin 0.01995815 0.0199483 FALSE
## 5:
                                                                      dead
## 6: 1244 3989.652 3993.655
                               Frog 0.70613136 0.7062234 FALSE
                                                                     right
##
      sentiment
## 1: positive
## 2: positive
## 3: positive
## 4:
      positive
## 5: negative
## 6: positive
# in this case, words can be repeated if they have more than one sentiment
aniNrcWordLonger <- animated %>%
  unnest_tokens(word, subtitle) %>%
  .[nrc, on = "word", nomatch = 0]
aniNrcWordLonger[sample(1:nrow(aniNrcWordLonger), size = 6)]
##
             start
                         end
                                movie startScale
                                                    endScale song
                                                                      word
        5
            71.593
                      76.509 Tangled 0.003233807 0.00352433 FALSE heavens
## 2: 1115 4356.446 4359.643 Sleeping 0.965943683 0.96500697 FALSE
                                                                     marry
## 3: 566 3658.826 3662.079
                                 Snow 0.725038192 0.72486177 FALSE
                                                                    grumpy
## 4: 783 3914.872 3917.625 Mermaid 0.832472201 0.83191786 FALSE
                                                                    lovely
## 5: 429 2073.747 2076.181 Aladdin 0.417609271 0.41750100 FALSE
                                                                       hit
## 6: 909 3801.600 3803.556
                               Moana 0.591670299 0.59172568 TRUE
                                                                      hide
##
      sentiment
## 1: positive
## 2:
          trust
## 3:
       sadness
## 4: positive
## 5: negative
## 6:
          fear
# this version is .001 seconds slower...
# aniNrcLineWider <- aniNrcWordLonger[, groupID := .GRP, by = list(n, movie)] \%%
#
   unique() %>%
   select(-c(word)) %>%
#
  pivot_wider(names_from = sentiment, values_from = 1, values_fn = length) %>%
   arrange(movie, start) %>%
   mutate(across(nrcSentiments, ~ifelse(is.na(.x), FALSE, TRUE))) %>%
   data.table()
# aniNrcLineWider[sample(1:nrow(aniNrcLineWider), size = 6)]
# creates a wider version of aniNrcWordLonger, but aggregates by line
# lines are identified with a `groupID` found by grouping by line number and movie
aniNrcLineWider <- aniNrcWordLonger[, groupID := .GRP, by = list(n, movie)] %>%
  unique() %>%
  .[, !c("word")] %>%
 pivot_wider(names_from = sentiment, values_from = 1, values_fn = length) %>%
  setorderv(c("movie", "start")) %>%
```

```
mutate(across(nrcSentiments, ~ifelse(is.na(.x), FALSE, TRUE))) %>%
 data.table()
## Note: Using an external vector in selections is ambiguous.
## i Use 'all_of(nrcSentiments)' instead of 'nrcSentiments' to silence this message.
## i See <https://tidyselect.r-lib.org/reference/faq-external-vector.html>.
## This message is displayed once per session.
aniNrcLineWider[sample(1:nrow(aniNrcLineWider), size = 6)]
                                               endScale song groupID anger fear
##
        start
                   end
                            movie startScale
## 1: 438.137 445.441
                         Sleeping 0.0910155 0.09196511 TRUE
                                                                   24 FALSE FALSE
## 2: 2810.891 2812.976
                             Frog 0.4947034 0.49442603 FALSE
                                                                 3827 FALSE FALSE
                          Mermaid 0.3153788 0.31545003 TRUE
## 3: 1530.362 1534.283
                                                                 3933 FALSE FALSE
## 4: 897.146 899.731
                             Frog 0.1514452 0.15121653 FALSE
                                                                 994 FALSE FALSE
## 5: 4565.978 4570.064
                             Frog 0.8095038 0.80962313 FALSE
                                                                 4447 FALSE FALSE
## 6: 4139.200 4143.400 Pocahontas 0.8564291 0.85632292 FALSE
                                                                 2114 TRUE TRUE
     negative sadness positive disgust surprise
                                                 joy anticipation trust
## 1:
        FALSE
               FALSE
                          TRUE
                                 FALSE
                                          FALSE TRUE
                                                              TRUE TRUE
## 2:
        FALSE
                 TRUE
                          TRUE
                                 FALSE
                                          FALSE TRUE
                                                             FALSE FALSE
## 3:
        FALSE
               FALSE
                          TRUE
                                 FALSE
                                          FALSE TRUE
                                                              TRUE TRUE
## 4:
        FALSE
                                          FALSE FALSE
                                                              TRUE FALSE
              FALSE
                         FALSE
                                 FALSE
## 5:
        FALSE
                                 FALSE
                                          FALSE TRUE
                                                             FALSE FALSE
               FALSE
                          TRUE
## 6:
        TRUE
                 TRUE
                          TRUE
                                 FALSE
                                          FALSE FALSE
                                                             FALSE FALSE
# I spent so much time on this but made it better above :(
# it also showcases the some of the tidyverse version of the nonsense above
# %>%
    left_join(aniNrcWordLonger %>%
               group_by(n, movie) %>%
#
               mutate(groupID = cur_group_id()) %>%
               select(-c(n, word, sentiment)) %>%
#
#
               unique(),
#
             by = "qroupID") \%>\%
#
  select(movie, n, start, end, nrcSentiments) %>%
  arrange(movie, n)
```

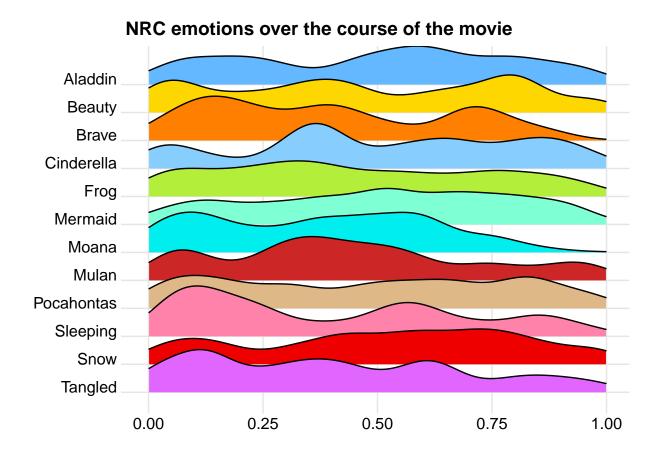
### Top ten most common spoken words by movie



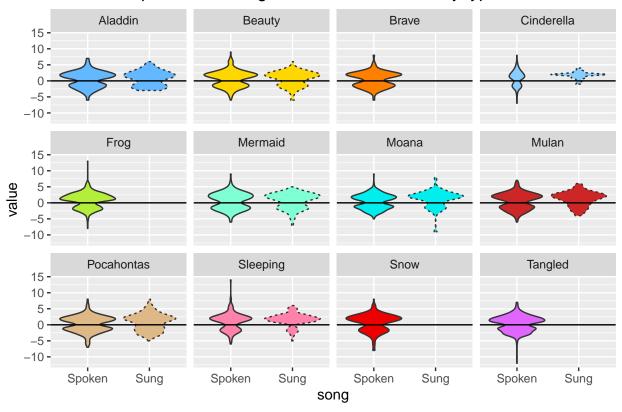
### Top ten most common sung words by movie



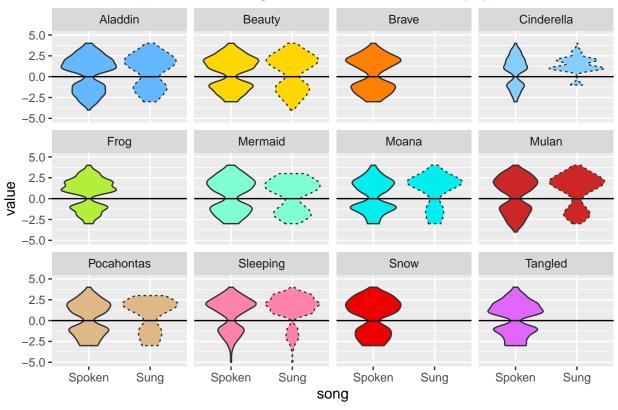
## Picking joint bandwidth of 0.0575



# Variation in positive and negative emotions of lines by type



# Variation in positive and negative emotions of words by type



##		movie	totalSent	timent					
##	1:	Frog		498					
##	2:	Sleeping		457					
##	3:	${\tt Cinderella}$	362						
##	4:	Beauty		355					
##	5:	Aladdin	304						
##	6:	Mulan	260						
##	7:	Mermaid	naid 214						
##	8:	Moana							
##	9:	Snow		192					
##	10:	Tangled		162					
##	11:	${\tt Pocahontas}$		160					
##	12:	Brave	113						
##		movie	negative	positive	totalSentiment				
##	1:	Cinderella	106	271	165				
##	2:	Frog	263	424	161				
##	3:	Sleeping	107	257	150				
##	4:	Tangled	185	324	139				
##	5:	Beauty	264	402	138				
##	6:	${\tt Pocahontas}$	221	358	137				
##	7:	Aladdin	270	397	127				
##	8:	Mulan	168	291	123				
##	9:	Moana	157	258	101				
##	10:				99				

##	11:	Mermaid	208 258		50					
##	12:	Snow	:	162 203		41				
##		movie	anger	anticipation	disgust	fear	joy	negative	positive	sadness
##	1:	Aladdin	107	207	97	193	223	253	475	119
##	2:	Beauty	118	208	53	191	191	289	381	122
##	3:	Brave	84	142	53	163	101	219	235	70
##	4:	${\tt Cinderella}$	36	221	36	54	208	153	373	81
##	5:	Frog	94	271	101	131	320	251	542	115
##	6:	Mermaid	70	177	57	119	174	178	334	120
##	7:	Moana	76	152	30	107	132	170	260	78
##	8:	Mulan	90	157	53	146	125	204	318	75
##	9:	${\tt Pocahontas}$	106	176	78	152	153	238	366	103
##	10:	Sleeping	47	187	40	83	227	119	425	56
##	11:	Snow	66	115	57	99	133	154	195	86
##	12:	Tangled	77	203	72	96	203	225	343	146
##		surprise to	rust to	otalSentiment	totalFee	eling				
##	1:	128	238	222		1312				
##	2:	126	205	92		1214				
##	3:	62	126	16		801				
##	4:	92	220	220		948				
##	5:	136	253	291		1421				
##	6:	107	133	156		957				
##	7:	56	128	90		759				
##	8:	81	206	114		933				
##	9:	104	216	128		1088				
	10:	124	168	306		932				
	11:	69	106	41		731				
##	12:	72	203	118		1072				

