

# Memory content

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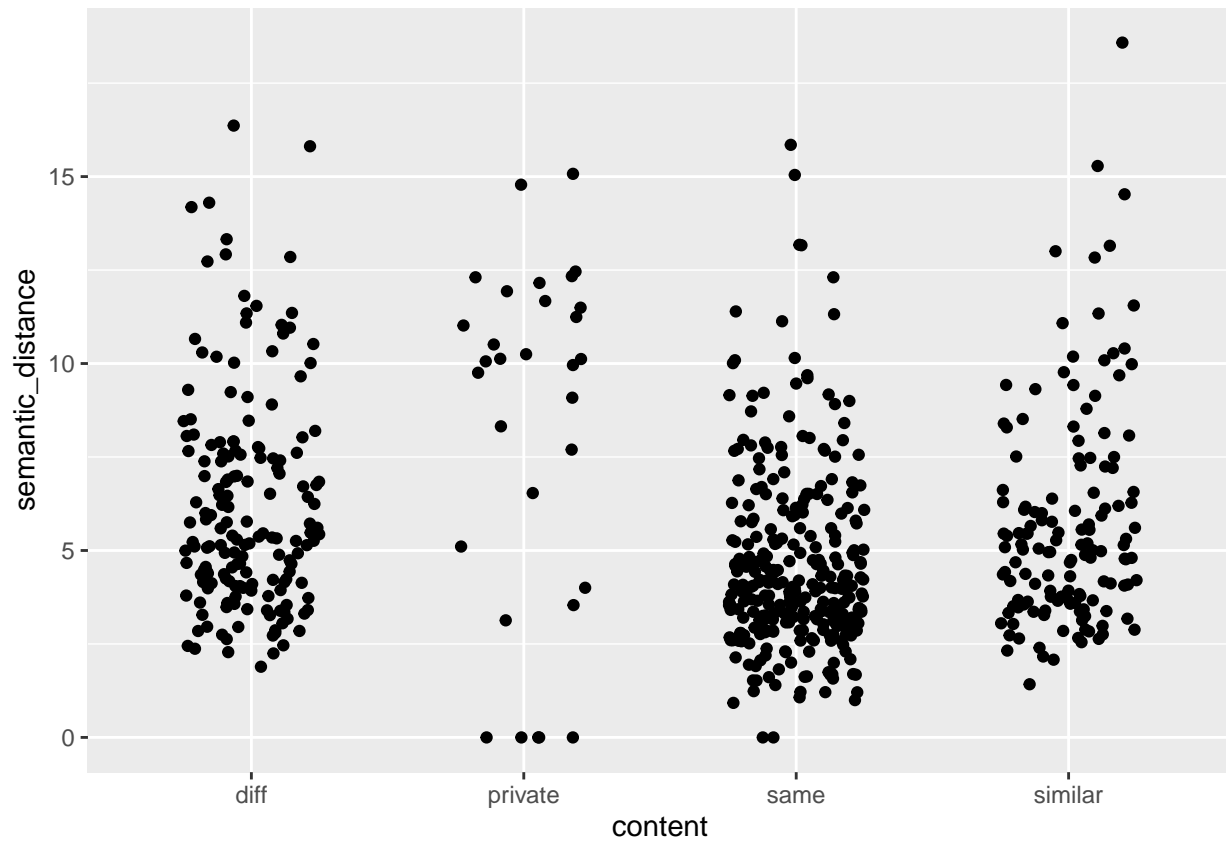
This notebook looks at memory content for pairs of original-cover memories, as predicted by perceptual similarity of the clips evoking the memories.

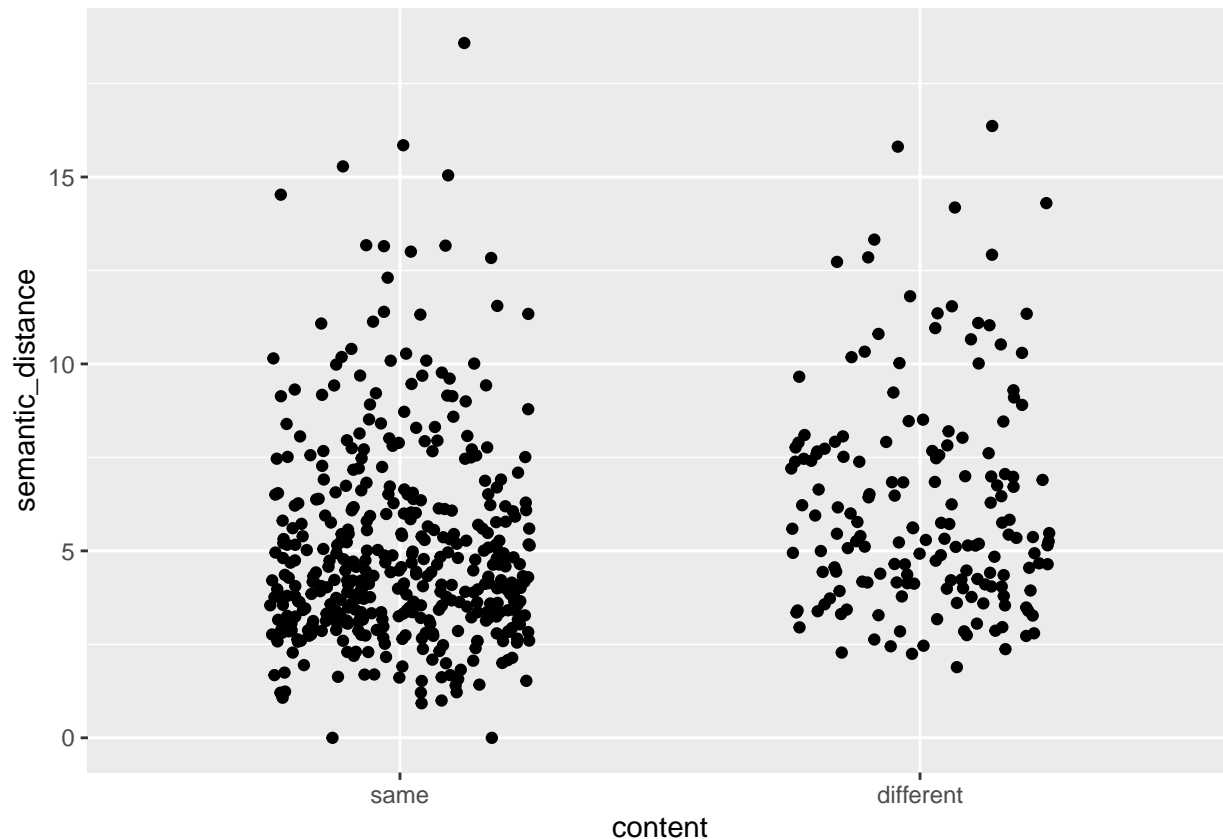
Load the data.

Load just the features and similarity for song-wise analysis.

Make same/different coding a factor. (Filter out private, lump similar in with same.)

Look at same/different coding vs semantic distance from sentence embeddings.





Predict semantic distance from perceptual similarity.

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: semantic_distance ~ mean_sim + (1 | internal_id) + (1 | song_id)
## Data: data_factor
##
## REML criterion at convergence: 2917.2
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.1880 -0.6487 -0.2439  0.4158  4.7441
##
## Random effects:
## Groups      Name                Variance Std.Dev.
## internal_id (Intercept) 0.5902     0.7683
## song_id      (Intercept) 0.0000     0.0000
## Residual                    7.2411     2.6909
## Number of obs: 598, groups:  internal_id, 79; song_id, 50
##
## Fixed effects:
##              Estimate Std. Error    df t value Pr(>|t|)
## (Intercept)   5.2672     0.1569 56.0995  33.576 <2e-16 ***
## mean_sim      -0.0219     0.1114 580.9586  -0.197  0.844
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
```

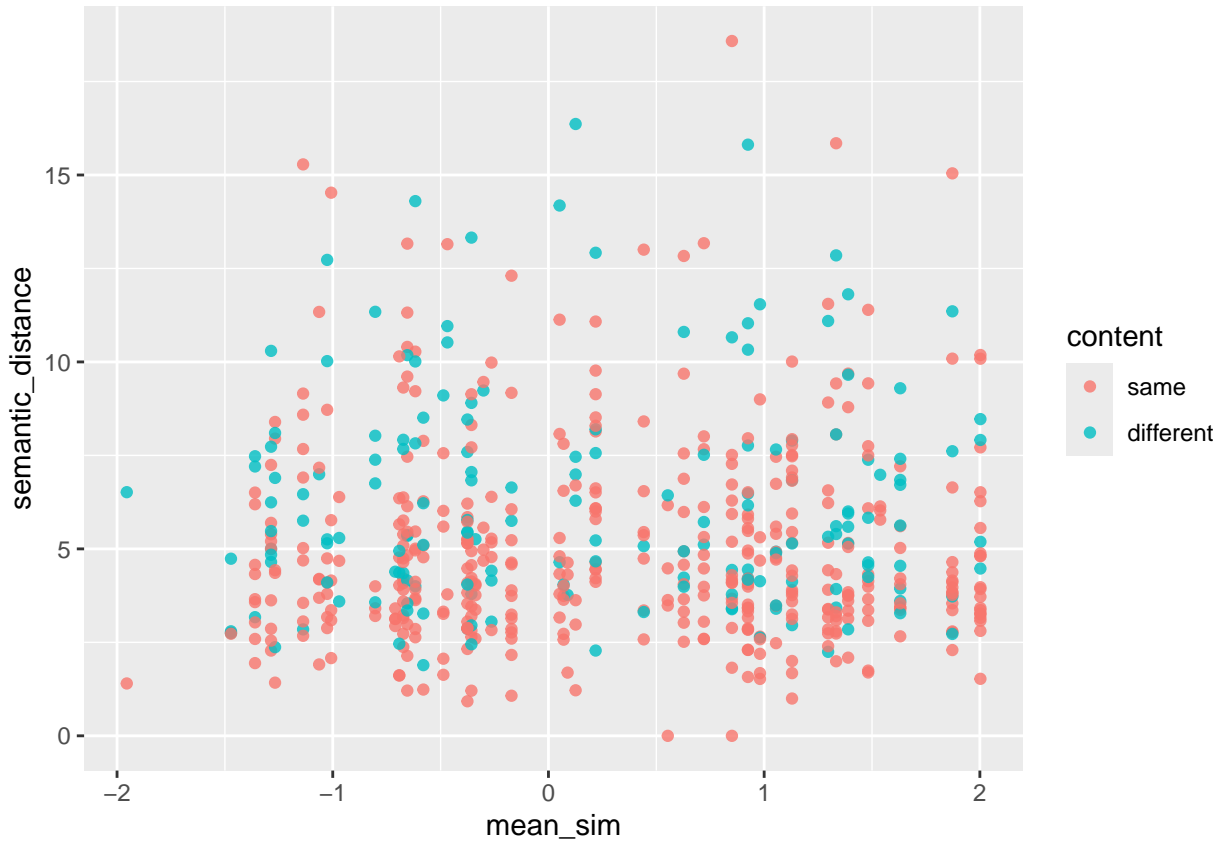
```

##          (Intr)
## mean_sim -0.215
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')

Predict content (same/different) from perceptual similarity.

## Generalized linear mixed model fit by maximum likelihood (Laplace
##   Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: content ~ mean_sim + (1 | internal_id) + (1 | song_id)
## Data: data_factor
##
##      AIC      BIC   logLik deviance df.resid
##  695.1    712.6   -343.5   687.1     594
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.2147 -0.6228 -0.4389  1.0171  2.7319
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## internal_id (Intercept) 0.8618   0.9283
## song_id      (Intercept) 0.0000   0.0000
## Number of obs: 598, groups: internal_id, 79; song_id, 50
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.10555    0.17114  -6.460 1.05e-10 ***
## mean_sim     -0.13664    0.09817  -1.392   0.164
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##          (Intr)
## mean_sim -0.132
## optimizer (Nelder_Mead) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')

```



Visualize.

Can we predict semantic distance just for same memories? (expect no)

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: semantic_distance ~ mean_sim + (1 | internal_id) + (1 | song_id)
## Data: .
##
## REML criterion at convergence: 2061.8
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.9963 -0.6378 -0.2730  0.4244  5.0312
##
## Random effects:
##  Groups      Name                Variance Std.Dev.
## internal_id (Intercept) 0.15932  0.3992
## song_id      (Intercept) 0.09654  0.3107
## Residual                    6.88808  2.6245
## Number of obs: 429, groups:  internal_id, 74; song_id, 50
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)  4.960573   0.154089 29.284616 32.193  <2e-16 ***
## mean_sim      0.003126   0.138054 34.835421  0.023   0.982
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
```

```

## Correlation of Fixed Effects:
##      (Intr)
## mean_sim -0.256

Can we predict semantic distance just for different memories? (expect yes??)

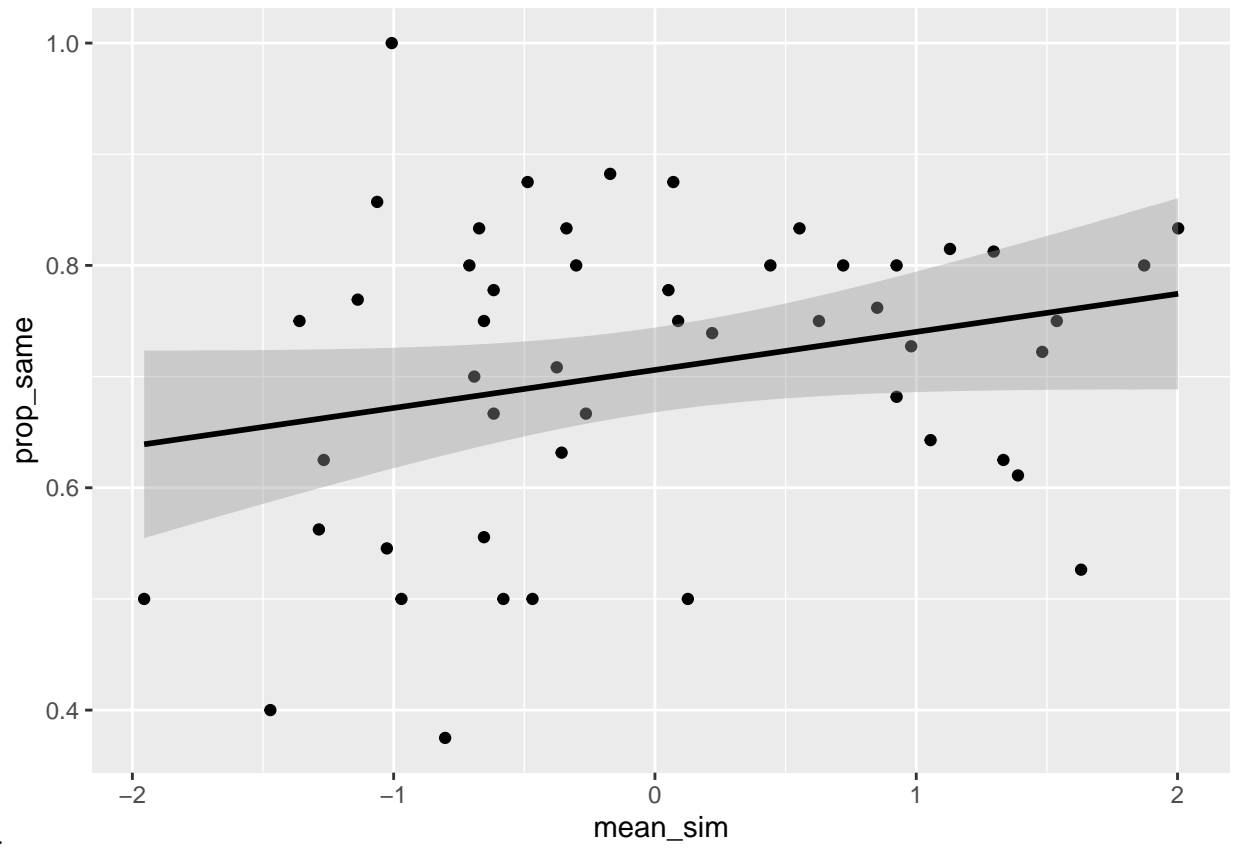
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: semantic_distance ~ mean_sim + (1 | internal_id) + (1 | song_id)
##   Data: .
##
## REML criterion at convergence: 832.9
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.4646 -0.7193 -0.2074  0.4978  3.3808
##
## Random effects:
##   Groups      Name      Variance Std.Dev.
## internal_id (Intercept) 1.463e+00 1.210e+00
## song_id      (Intercept) 1.102e-09 3.319e-05
## Residual                7.020e+00 2.650e+00
## Number of obs: 169, groups:  internal_id, 49; song_id, 49
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)   6.12401    0.29155   43.36182  21.005   <2e-16 ***
## mean_sim      0.06805    0.21015  162.94987   0.324   0.746
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##      (Intr)
## mean_sim -0.156
## optimizer (nloptwrap) convergence code: 0 (OK)
## boundary (singular) fit: see help('isSingular')

Compute proportion of same or similar memories per song.

##
## Call:
## lm(formula = prop_same ~ mean_sim, data = data_by_song)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.30351 -0.09942  0.02410  0.08765  0.32848
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   0.70599    0.01895  37.249   <2e-16 ***
## mean_sim      0.03422    0.01915   1.787   0.0802 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.134 on 48 degrees of freedom
## Multiple R-squared:  0.0624, Adjusted R-squared:  0.04287

```

## F-statistic: 3.194 on 1 and 48 DF, p-value: 0.0802



Visualize.