

Variance partitioning

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2025-01-13

This notebook uses the musical and emotional features and performs variance partitioning to understand how these sets of features work together to explain similarity.

Load the data.

Keep the musical and emotional features.

With help from: <https://www.davidzeleny.net/anadat-r/doku.php/en:varpart>

Venn diagram parts:

- a: variance explained by musical features only
- b: variance explained by musical and emotional features together
- c: variance explained by emotional features only
- d: unexplained variance

For the musical features, we'll allow key, voice type, and MFCCs to interact because that was the best musical features model (see `musical_features.Rmd`).

Calculate manually

[a+b+c]

```
##
## Call:
## rda(formula = data$mean_sim ~ tempo_difference + key_difference *      voice_type_difference * mfccs,
##
## Partitioning of variance:
##           Inertia Proportion
## Total           1.0000      1.0000
## Constrained      0.6426      0.6426
## Unconstrained    0.3574      0.3574
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##           RDA1      PC1
## Eigenvalue           0.6426 0.3574
## Proportion Explained  0.6426 0.3574
## Cumulative Proportion 0.6426 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
##           RDA1
## Eigenvalue           0.6426
## Proportion Explained  1.0000
## Cumulative Proportion 1.0000
```

Is this consistent with just a simple linear model?

Yes, the explained variance is 0.5821, which is similar to the “constrained” variance in the RDA model. For RDA, it seems like “unconstrained” variance is [d].

Model for musical features only [a+b]

```
##
## Call:
## rda(formula = data$mean_sim ~ tempo_difference + key_difference *      voice_type_difference * mfccs,
##
## Partitioning of variance:
##           Inertia Proportion
## Total           1.0000      1.0000
## Constrained      0.5607      0.5607
## Unconstrained    0.4393      0.4393
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##           RDA1      PC1
## Eigenvalue           0.5607 0.4393
## Proportion Explained  0.5607 0.4393
## Cumulative Proportion 0.5607 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
##           RDA1
```

```

## Eigenvalue          0.5607
## Proportion Explained 1.0000
## Cumulative Proportion 1.0000

Model for emotional features only [b+c]

##
## Call:
## rda(formula = data$mean_sim ~ mean_valence_difference + mean_arousal_difference,      data = data)
##
## Partitioning of variance:
##           Inertia Proportion
## Total           1.0000      1.0000
## Constrained      0.3944      0.3944
## Unconstrained    0.6056      0.6056
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##           RDA1      PC1
## Eigenvalue      0.3944 0.6056
## Proportion Explained 0.3944 0.6056
## Cumulative Proportion 0.3944 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
##           RDA1
## Eigenvalue      0.3944
## Proportion Explained 1.0000
## Cumulative Proportion 1.0000

```

Extract the R^2 values.

Subtract to find individual fractions.

- variance explained by musical features only (a): 20.58%
- shared variance explained (b): 29.74%
- variance explained by emotional features only (c): 7.89%
- unexplained variance (d): 41.79%

Musical features

- a: variance explained by tempo only
- b: shared variance explained by the musical features
- c: variance explained by key, voice type, and MFCCs only
- d: unexplained variance

[a+b+c]

```
##
## Call:
## rda(formula = data$mean_sim ~ tempo_difference + key_difference *      voice_type_difference * mfccs,
##
## Partitioning of variance:
##              Inertia Proportion
## Total          1.0000      1.0000
## Constrained    0.5607      0.5607
## Unconstrained  0.4393      0.4393
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##              RDA1      PC1
## Eigenvalue          0.5607 0.4393
## Proportion Explained 0.5607 0.4393
## Cumulative Proportion 0.5607 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
##              RDA1
## Eigenvalue          0.5607
## Proportion Explained 1.0000
## Cumulative Proportion 1.0000
```

[a+b]

```
##
## Call:
## rda(formula = data$mean_sim ~ tempo_difference, data = data)
##
## Partitioning of variance:
##              Inertia Proportion
## Total          1.0000      1.0000
## Constrained    0.3386      0.3386
## Unconstrained  0.6614      0.6614
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##              RDA1      PC1
## Eigenvalue          0.3386 0.6614
## Proportion Explained 0.3386 0.6614
## Cumulative Proportion 0.3386 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
```

```

##                                RDA1
## Eigenvalue                    0.3386
## Proportion Explained    1.0000
## Cumulative Proportion 1.0000

[b+c]

##
## Call:
## rda(formula = data$mean_sim ~ key_difference * voice_type_difference *      mfccs_dist, data = data)
##
## Partitioning of variance:
##              Inertia Proportion
## Total          1.0000      1.0000
## Constrained    0.3174      0.3174
## Unconstrained  0.6826      0.6826
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##              RDA1      PC1
## Eigenvalue      0.3174 0.6826
## Proportion Explained 0.3174 0.6826
## Cumulative Proportion 0.3174 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
##              RDA1
## Eigenvalue      0.3174
## Proportion Explained 1.0000
## Cumulative Proportion 1.0000

```

Extract the R^2 values.

Subtract to find individual fractions.

- variance explained by tempo only (a): 26.28%
- shared variance (b): 6.61%
- variance explained by key, voice type, and timbre only (c): 17.42%
- unexplained variance (d): 49.69%

Emotional features

- a: variance explained by valence only
- b: shared variance explained by the emotional features
- c: variance explained by arousal only
- d: unexplained variance

[a+b+c]

```
##
## Call:
## rda(formula = data$mean_sim ~ mean_valence_difference + mean_arousal_difference,      data = data)
##
## Partitioning of variance:
##              Inertia Proportion
## Total          1.0000      1.0000
## Constrained     0.3944      0.3944
## Unconstrained   0.6056      0.6056
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##              RDA1      PC1
## Eigenvalue          0.3944 0.6056
## Proportion Explained 0.3944 0.6056
## Cumulative Proportion 0.3944 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
##              RDA1
## Eigenvalue          0.3944
## Proportion Explained 1.0000
## Cumulative Proportion 1.0000
```

[a+b]

```
##
## Call:
## rda(formula = data$mean_sim ~ mean_valence_difference, data = data)
##
## Partitioning of variance:
##              Inertia Proportion
## Total          1.0000      1.0000
## Constrained     0.3045      0.3045
## Unconstrained   0.6955      0.6955
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##              RDA1      PC1
## Eigenvalue          0.3045 0.6955
## Proportion Explained 0.3045 0.6955
## Cumulative Proportion 0.3045 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
```

```
##                                RDA1
## Eigenvalue                    0.3045
## Proportion Explained    1.0000
## Cumulative Proportion 1.0000

[b+c]

##
## Call:
## rda(formula = data$mean_sim ~ mean_arousal_difference, data = data)
##
## Partitioning of variance:
##              Inertia Proportion
## Total          1.0000      1.0000
## Constrained    0.2029      0.2029
## Unconstrained  0.7971      0.7971
##
## Eigenvalues, and their contribution to the variance
##
## Importance of components:
##              RDA1      PC1
## Eigenvalue      0.2029 0.7971
## Proportion Explained 0.2029 0.7971
## Cumulative Proportion 0.2029 1.0000
##
## Accumulated constrained eigenvalues
## Importance of components:
##              RDA1
## Eigenvalue      0.2029
## Proportion Explained 1.0000
## Cumulative Proportion 1.0000
```

Extract the R^2 values.

Subtract to find individual fractions.

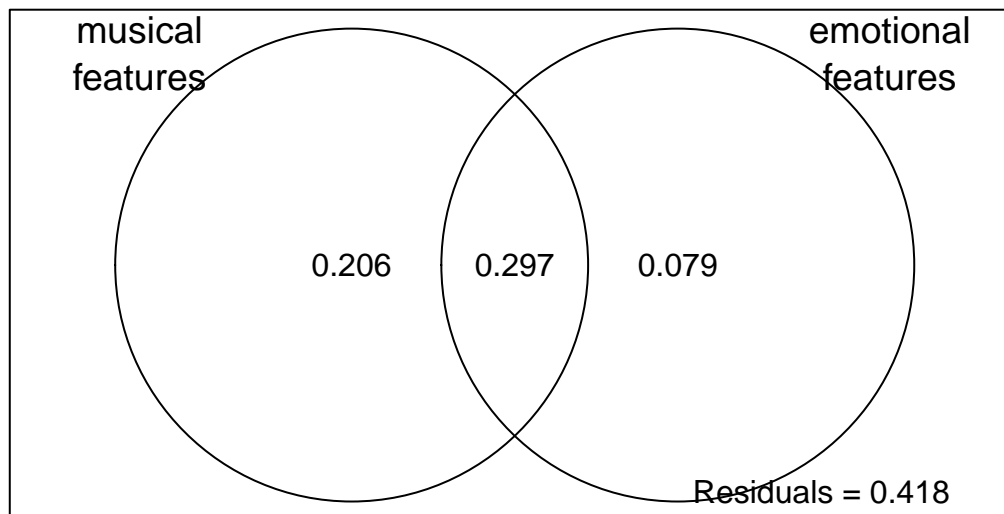
- variance explained by valence only (a): 18.51%
- shared variance by the emotional features (b): 10.92%
- variance explained by arousal only (c): 8.20%
- unexplained variance (d): 62.37%

Using varpart

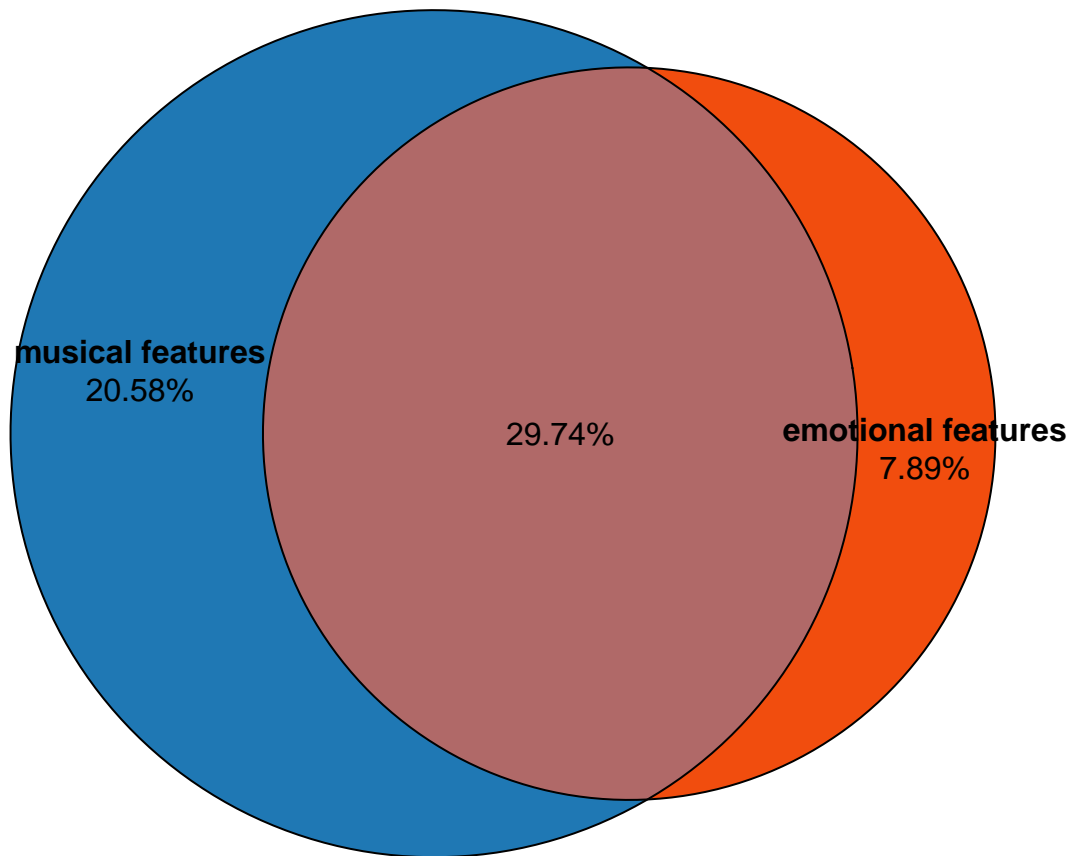
Musical and emotional features as a set

```
##
## Partition of variance in RDA
##
## Call: varpart(Y = data$mean_sim, X = ~tempo_difference + key_difference
## * voice_type_difference * mfccs_dist, ~mean_valence_difference +
## mean_arousal_difference, data = data)
##
## Explanatory tables:
## X1: ~tempo_difference + key_difference * voice_type_difference * mfccs_dist
## X2: ~mean_valence_difference + mean_arousal_difference
##
## No. of explanatory tables: 2
## Total variation (SS): 69
##           Variance: 1
## No. of observations: 70
##
## Partition table:
##           Df R.squared Adj.R.squared Testable
## [a+c] = X1      8  0.56074      0.50313     TRUE
## [b+c] = X2      2  0.39437      0.37629     TRUE
## [a+b+c] = X1+X2 10  0.64264      0.58207     TRUE
## Individual fractions
## [a] = X1|X2      8           0.20578     TRUE
## [b] = X2|X1      2           0.07894     TRUE
## [c]              0           0.29736    FALSE
## [d] = Residuals           0.41793    FALSE
## ---
## Use function 'rda' to test significance of fractions of interest
```

Plot the results with a Venn diagram



Plot an area-proportional Venn diagram



Separating into three sets (tempo, all other musical features, emotional features)

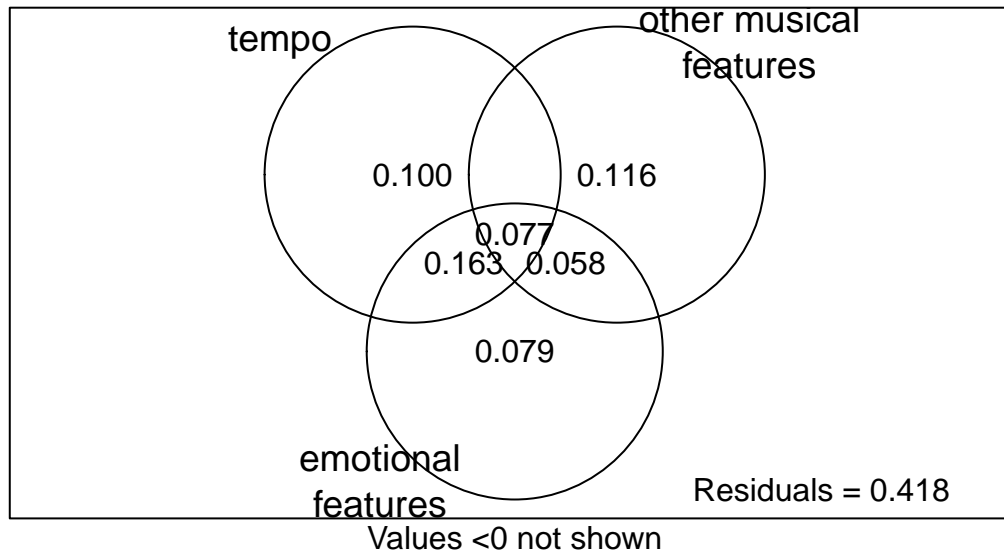
```
##
## Partition of variance in RDA
##
## Call: varpart(Y = data$mean_sim, X = ~tempo_difference, ~key_difference
## * voice_type_difference * mfccs_dist, ~mean_valence_difference +
## mean_arousal_difference, data = data)
##
## Explanatory tables:
## X1: ~tempo_difference
## X2: ~key_difference * voice_type_difference * mfccs_dist
## X3: ~mean_valence_difference + mean_arousal_difference
##
## No. of explanatory tables: 3
## Total variation (SS): 69
##           Variance: 1
## No. of observations: 70
##
## Partition table:
##
```

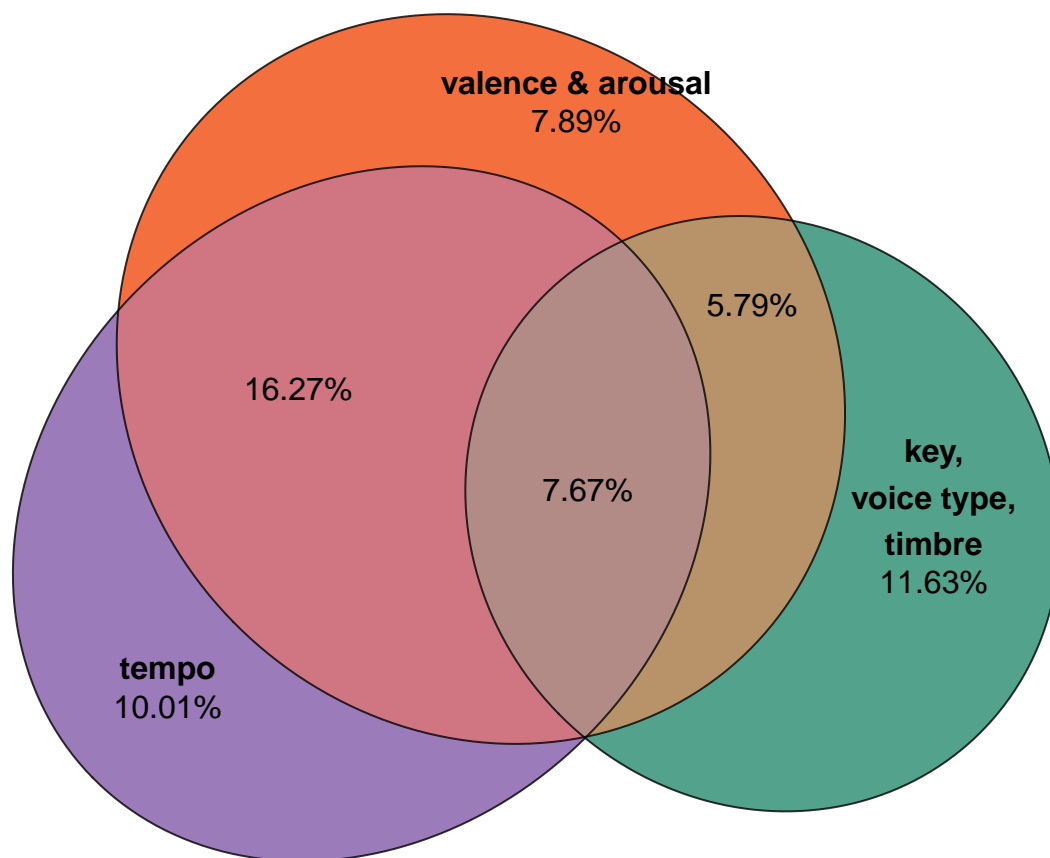
	Df	R.square	Adj.R.square	Testable
## [a+d+f+g] = X1	1	0.33864	0.32892	TRUE
## [b+d+e+g] = X2	7	0.31736	0.24029	TRUE
## [c+e+f+g] = X3	2	0.39437	0.37629	TRUE
## [a+b+d+e+f+g] = X1+X2	8	0.56074	0.50313	TRUE
## [a+c+d+e+f+g] = X1+X3	3	0.48900	0.46578	TRUE
## [b+c+d+e+f+g] = X2+X3	9	0.54953	0.48195	TRUE

```

## [a+b+c+d+e+f+g] = All 10 0.64264 0.58207 TRUE
## Individual fractions
## [a] = X1 | X2+X3 1 0.10012 TRUE
## [b] = X2 | X1+X3 7 0.11629 TRUE
## [c] = X3 | X1+X2 2 0.07894 TRUE
## [d] 0 -0.01063 FALSE
## [e] 0 0.05793 FALSE
## [f] 0 0.16273 FALSE
## [g] 0 0.07670 FALSE
## [h] = Residuals 0.41793 FALSE
## Controlling 1 table X
## [a+d] = X1 | X3 1 0.08949 TRUE
## [a+f] = X1 | X2 1 0.26285 TRUE
## [b+d] = X2 | X3 7 0.10566 TRUE
## [b+e] = X2 | X1 7 0.17422 TRUE
## [c+e] = X3 | X1 2 0.13686 TRUE
## [c+f] = X3 | X2 2 0.24167 TRUE
## ---
## Use function 'rda' to test significance of fractions of interest

```



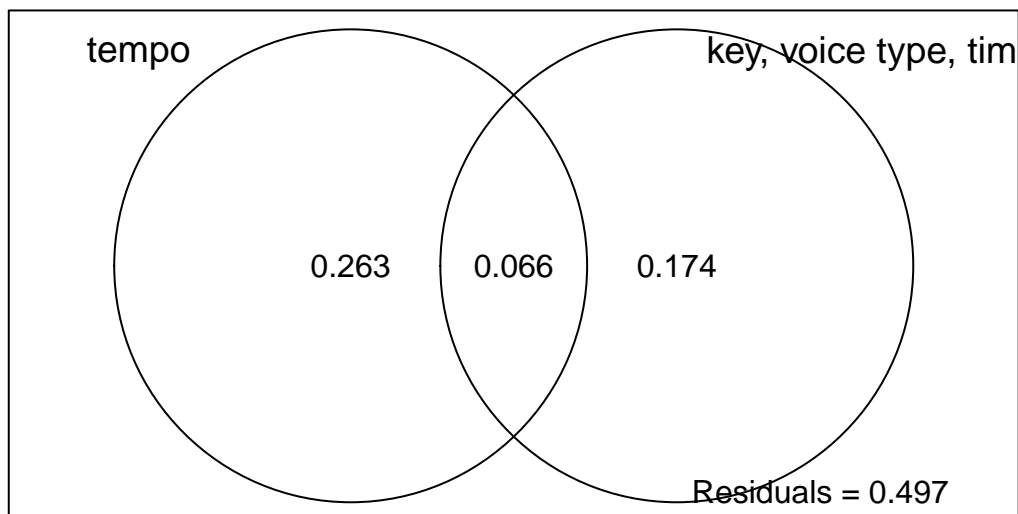


Venn diagrams for musical features only and emotional features only.

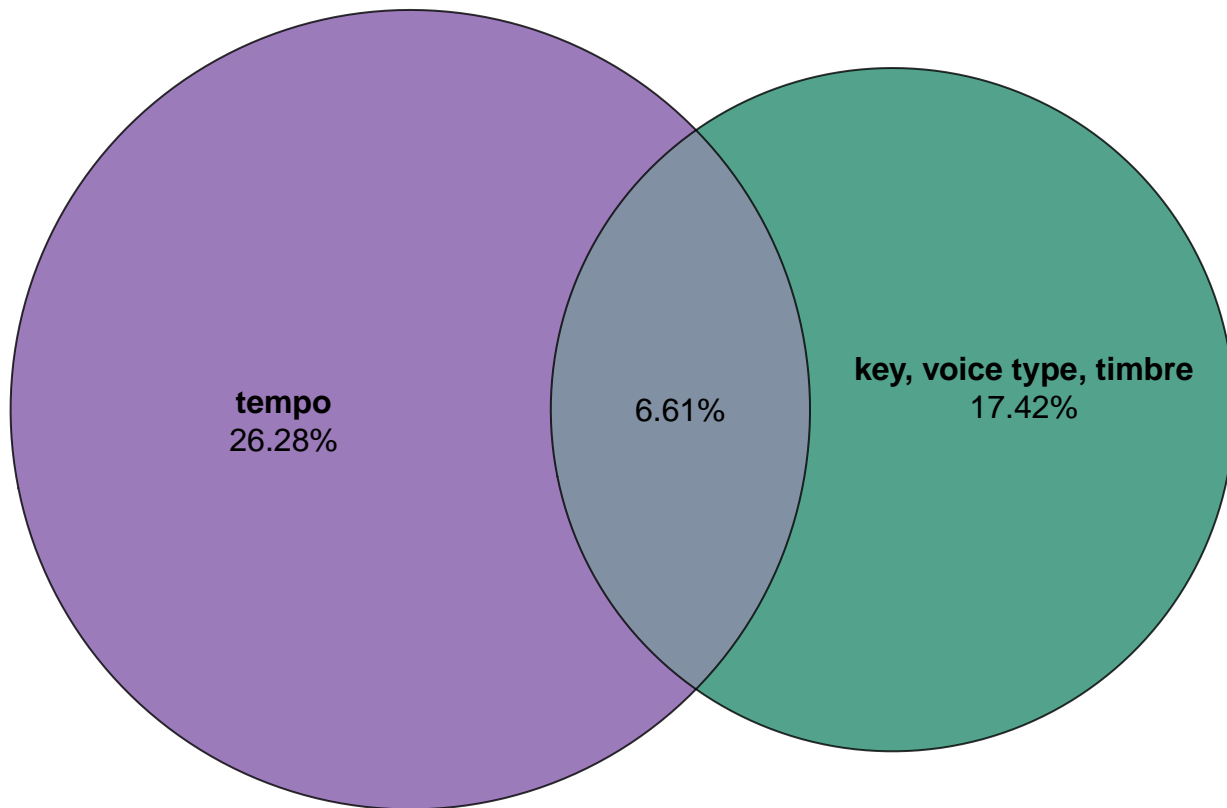
Musical features only

```
##
## Partition of variance in RDA
##
## Call: varpart(Y = data$mean_sim, X = ~tempo_difference, ~key_difference
## * voice_type_difference * mfccs_dist, data = data)
##
## Explanatory tables:
## X1: ~tempo_difference
## X2: ~key_difference * voice_type_difference * mfccs_dist
##
## No. of explanatory tables: 2
## Total variation (SS): 69
##          Variance: 1
## No. of observations: 70
##
## Partition table:
##
##      Df R.squared Adj.R.squared Testable
## [a+c] = X1      1  0.33864      0.32892   TRUE
## [b+c] = X2      7  0.31736      0.24029   TRUE
## [a+b+c] = X1+X2  8  0.56074      0.50313   TRUE
## Individual fractions
## [a] = X1|X2      1           0.26285   TRUE
## [b] = X2|X1      7           0.17422   TRUE
## [c]              0           0.06607   FALSE
## [d] = Residuals           0.49687   FALSE
## ---
## Use function 'rda' to test significance of fractions of interest
```

Plot the results with a Venn diagram



Plot an area-proportional Venn diagram

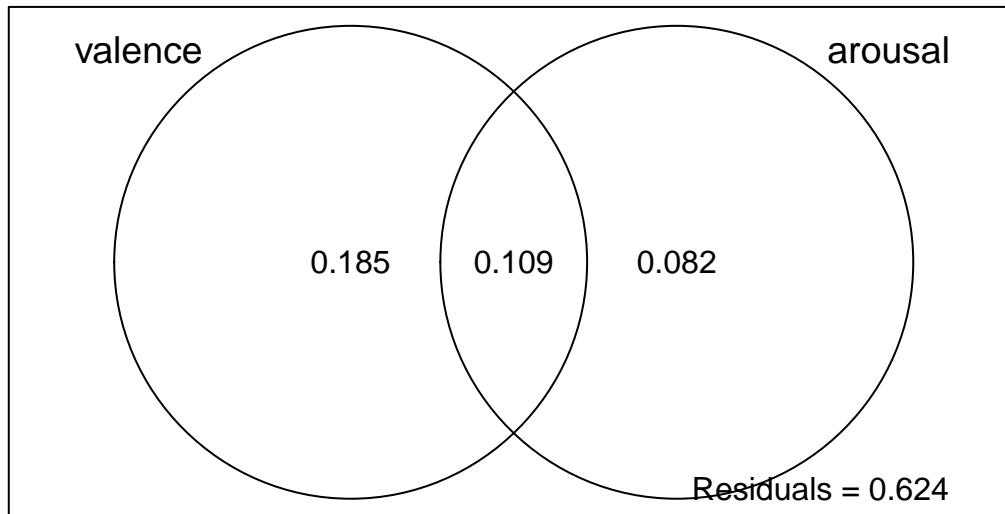


Emotional features only

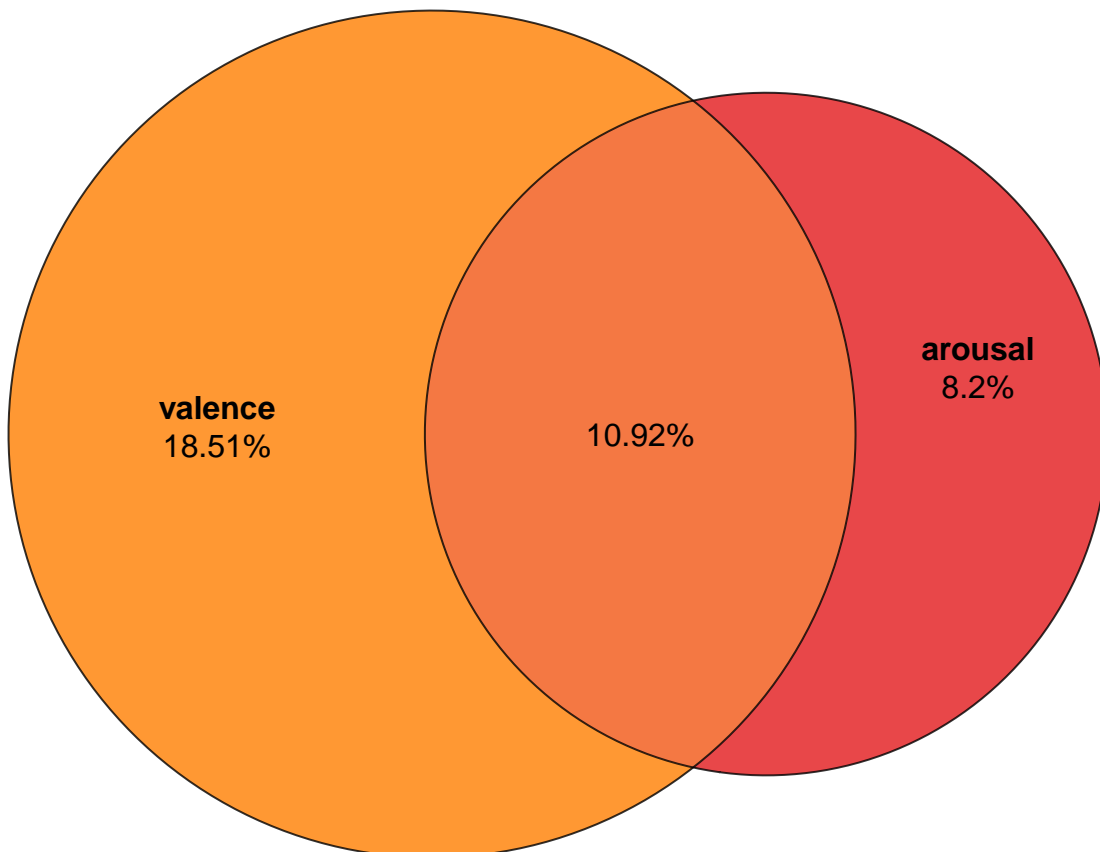
```
##
## Partition of variance in RDA
##
## Call: varpart(Y = data$mean_sim, X = ~mean_valence_difference,
## ~mean_arousal_difference, data = data)
##
## Explanatory tables:
## X1: ~mean_valence_difference
## X2: ~mean_arousal_difference
##
## No. of explanatory tables: 2
## Total variation (SS): 69
##          Variance: 1
## No. of observations: 70
##
## Partition table:
##           Df R.squared Adj.R.squared Testable
## [a+c] = X1      1  0.30455      0.29432    TRUE
## [b+c] = X2      1  0.20288      0.19116    TRUE
## [a+b+c] = X1+X2  2  0.39437      0.37629    TRUE
## Individual fractions
## [a] = X1|X2      1              0.18513    TRUE
## [b] = X2|X1      1              0.08197    TRUE
## [c]              0              0.10919    FALSE
## [d] = Residuals              0.62371    FALSE
## ---
```

```
## Use function 'rda' to test significance of fractions of interest
```

Plot the results with a Venn diagram



Plot an area-proportional Venn diagram



Can we do this with valence and arousal separately in a four-way model?

```
##
## Partition of variance in RDA
##
## Call: varpart(Y = data$mean_sim, X = ~tempo_difference, ~key_difference
## * voice_type_difference * mfccs_dist, ~mean_valence_difference,
## ~mean_arousal_difference, data = data)
##
## Explanatory tables:
## X1: ~tempo_difference
## X2: ~key_difference * voice_type_difference * mfccs_dist
## X3: ~mean_valence_difference
## X4: ~mean_arousal_difference
##
## No. of explanatory tables: 4
## Total variation (SS): 69
## Variance: 1
## No. of observations: 70
##
## Partition table:
```

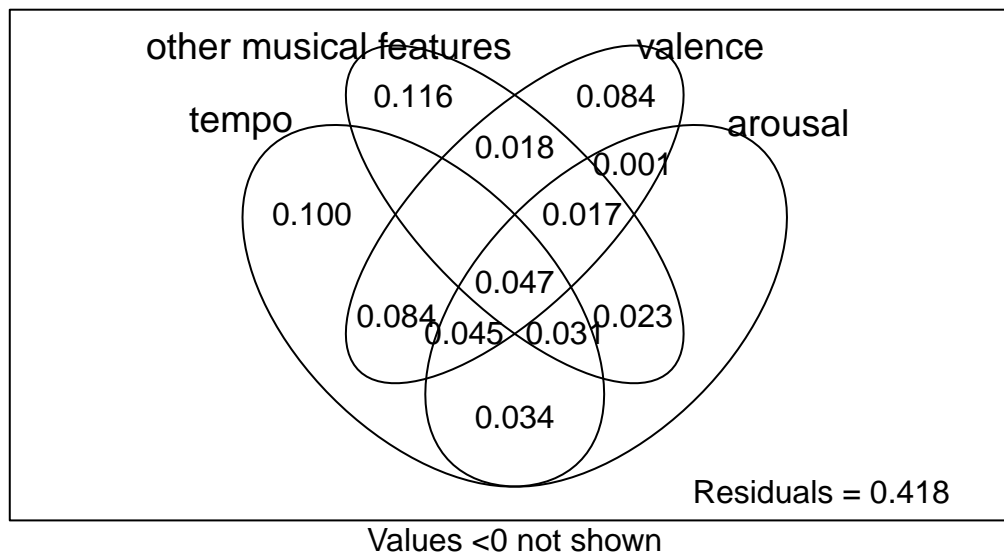
	Df	R.square	Adj.R.square	Testable
## [aeghklno] = X1	1	0.33864	0.32892	TRUE
## [befiklmo] = X2	7	0.31736	0.24029	TRUE
## [cfgjlmno] = X3	1	0.30455	0.29432	TRUE
## [dhijkmno] = X4	1	0.20288	0.19116	TRUE
## [abefghijklmno] = X1+X2	8	0.56074	0.50313	TRUE
## [acefghijklmno] = X1+X3	2	0.46385	0.44784	TRUE
## [adehijklmno] = X1+X4	2	0.38248	0.36404	TRUE
## [bcefgijklmno] = X2+X3	8	0.51711	0.45378	TRUE
## [bdefhijklmno] = X2+X4	8	0.39341	0.31386	TRUE
## [cdfghijklmno] = X3+X4	2	0.39437	0.37629	TRUE
## [abcefgijklmno] = X1+X2+X3	9	0.64124	0.58743	TRUE
## [abdefghijklmno] = X1+X2+X4	9	0.56386	0.49844	TRUE
## [acdefghijklmno] = X1+X3+X4	3	0.48900	0.46578	TRUE
## [bcdefghijklmno] = X2+X3+X4	9	0.54953	0.48195	TRUE
## [abcdefghijklmno] = All	10	0.64264	0.58207	TRUE
## Individual fractions				
## [a] = X1 X2+X3+X4	1		0.10012	TRUE
## [b] = X2 X1+X3+X4	7		0.11629	TRUE
## [c] = X3 X1+X2+X4	1		0.08363	TRUE
## [d] = X4 X1+X2+X3	1		-0.00536	TRUE
## [e]	0		-0.01063	FALSE
## [f]	0		0.01811	FALSE
## [g]	0		0.08447	FALSE
## [h]	0		0.03353	FALSE
## [i]	0		0.02329	FALSE
## [j]	0		0.00067	FALSE
## [k]	0		0.03051	FALSE
## [l]	0		-0.00107	FALSE
## [m]	0		0.01652	FALSE
## [n]	0		0.04473	FALSE
## [o]	0		0.04727	FALSE
## [p] = Residuals	0		0.41793	FALSE

```

## Controlling 2 tables X
## [ae] = X1 | X3+X4          1          0.08949    TRUE
## [ag] = X1 | X2+X4          1          0.18459    TRUE
## [ah] = X1 | X2+X3          1          0.13364    TRUE
## [be] = X2 | X3+X4          7          0.10566    TRUE
## [bf] = X2 | X1+X4          7          0.13440    TRUE
## [bi] = X2 | X1+X3          7          0.13959    TRUE
## [cf] = X3 | X1+X4          1          0.10174    TRUE
## [cg] = X3 | X2+X4          1          0.16810    TRUE
## [cj] = X3 | X1+X2          1          0.08429    TRUE
## [dh] = X4 | X2+X3          1          0.02817    TRUE
## [di] = X4 | X1+X3          1          0.01794    TRUE
## [dj] = X4 | X1+X2          1         -0.00469    TRUE
## Controlling 1 table X
## [aghn] = X1 | X2           1          0.26285    TRUE
## [aehk] = X1 | X3           1          0.15352    TRUE
## [aegl] = X1 | X4           1          0.17288    TRUE
## [bfim] = X2 | X1           7          0.17422    TRUE
## [beik] = X2 | X3           7          0.15946    TRUE
## [befl] = X2 | X4           7          0.12270    TRUE
## [cfjm] = X3 | X1           1          0.11892    TRUE
## [cgjn] = X3 | X2           1          0.21349    TRUE
## [cfgl] = X3 | X4           1          0.18513    TRUE
## [dijm] = X4 | X1           1          0.03513    TRUE
## [dhjn] = X4 | X2           1          0.07357    TRUE
## [dhik] = X4 | X3           1          0.08197    TRUE
## ---
## Use function 'rda' to test significance of fractions of interest

```

Plot the results with a Venn diagram



Try to plot an area-proportional Venn diagram

