

Cultural distances: Wikipedia data

Contents

Introduction	1
Load libraries	1
All domains	3
Load data	3
LMER models	6
Tests within domains	13
Load data	14
LMER models	14
Mantel tests	22
Data prep	22
Tests	24
Are the relevant D-place features most predictive?	26
References	26

Introduction

We compare cultural distances between societies with linguistic similarities between societies, controlling for shared history in two ways.

The first test uses mixed effects modelling. The pairing of the language family of each language (according to Glottolog) is used as a random effect. That means that the model can capture the likelihood that two languages from the Indo-European language family will be more similar to each other than two languages from different language families. The same is done with geographic area according to Autotyp.

The second test controls for history using distances from a phylogenetic tree. The tree comes from Bouckaert et al. (2012). Patristic distances between languages are used as a measure of historical distance between societies in a Mantel test. Note that the Mantel test assumes a strict distance metric, which is not necessarily the case with this data, but there are few other ways to deal with continuous pairwise distances.

Load libraries

```
library(ape)
library(ecodist)
library(lme4)
library(sjPlot)
library(ggplot2)
library(igraph)
library(lattice)
```

Parameters:

```
datasetName = "wikipedia-main"  
lingDistancesFile = "../data/FAIR/main-data-nel-wikipedia-k100-by-language-pair.csv"  
lingDistancesByDomainFile = "../results/EA_distances/wikipedia_All_Domains_with_ling.csv"  
# (generated by ../processing/combineCultAndLingDistances.R)
```

All domains

Load data

Read the cultural distances:

```
cult = read.csv("../results/EA_distances/CulturalDistances_Long.csv", stringsAsFactors = F)
names(cult) = c("l1", "l2", "cult.dist")
cultLangs = unique(c(cult$Var1, cult$Var2))
```

Add language family:

```
l = read.csv("../data/FAIR_langauges_glottol_xdid.csv", stringsAsFactors = F)
g = read.csv("../data/glottolog-languoid.csv/languoid.csv", stringsAsFactors = F)
l$family = g[match(l$glotto, g$id),]$family_pk
l$family = g[match(l$family, g$pk),]$name
```

Read the semantic distances

```
ling = read.csv(lingDistancesFile, stringsAsFactors = F)
```

There are very few possible comparisons for Slovenian and Northern Sami, so we'll remove these:

```
ling = ling[!(ling$l1=="se" || ling$l2 == "se"),]
ling = ling[!(ling$l1=="sl" || ling$l2 == "sl"),]
```

Combine the linguistic and cultural distances

```
cult$l1.iso2 = l[match(cult$l1, l$Language2),]$iso2
cult$l2.iso2 = l[match(cult$l2, l$Language2),]$iso2

fairisos = unique(c(ling$l1, ling$l2))
cultisos = unique(c(cult$l1.iso2, cult$l2.iso2))

cult = cult[(cult$l1.iso2 %in% fairisos) & (cult$l2.iso2 %in% fairisos),]
ling = ling[(ling$l1 %in% cultisos) & (ling$l2 %in% cultisos),]

matches = sapply(1:nrow(ling), function(i){
  which(cult$l1.iso2==ling$l1[i] & cult$l2.iso2==ling$l2[i])
})

ling$cult.dist = cult[matches,]$cult.dist
ling$cult.dist.center = scale(ling$cult.dist)
cdc.s = attr(ling$cult.dist.center, "scaled:scale")
cdc.c = attr(ling$cult.dist.center, "scaled:center")
ling$cult.dist.center = as.numeric(ling$cult.dist.center)

ling$family1 = l[match(ling$l1, l$iso2),]$family
ling$family2 = l[match(ling$l2, l$iso2),]$family
ling$area1 = l[match(ling$l1, l$iso2),]$autotyp.area
ling$area2 = l[match(ling$l2, l$iso2),]$autotyp.area

fgroup = cbind(ling$family1, ling$family2)
fgroup = apply(fgroup, 1, sort)
ling$family.group = apply(fgroup, 2, paste, collapse=":")
agroup = cbind(ling$area1, ling$area2)
```

```

agroup = apply(agroup,1,sort)
ling$area.group = apply(agroup,2,paste,collapse=":")

ling$rho.center = scale(ling$local_alignment)

```

Each observation is now associated with a language family pair:

```
head(ling[,c("l1","l2","local_alignment","family.group"])]))
```

```

##      l1 l2 local_alignment      family.group
## 3   ab he      0.07566733 Abkhaz-Adyge:Afro-Asiatic
## 5   ab zh      0.02225169 Abkhaz-Adyge:Sino-Tibetan
## 38  ba ab      0.07330162      Abkhaz-Adyge:Turkic
## 42  ba cv      0.18792509      Turkic:Turkic
## 44  ba he      0.17701891      Afro-Asiatic:Turkic
## 45  ba ja      0.13254527      Japonic:Turkic

```

And the same is true for area:

```
tail(ling[,c("l1","l2","local_alignment","area.group"])]))
```

```

##      l1 l2 local_alignment      area.group
## 2527 xal ja      0.02832668      Inner Asia:N Coast Asia
## 2530 xal ko      0.04879589      Inner Asia:N Coast Asia
## 2532 xal ml      0.07500293      Indic:Inner Asia
## 2533 xal ta      0.06311429      Indic:Inner Asia
## 2534 xal te      0.05080388      Indic:Inner Asia
## 2535 xal zh      0.02895876      Inner Asia:Southeast Asia

```

Number of observations:

```

# Number of datapoints:
nrow(ling)

```

```
## [1] 733
```

```

# Number of unique languages:
length(unique(unlist(ling[,c("l1","l2")]])))

```

```
## [1] 40
```

```

# Number of unique language families:
uniqueFamilies = unique(unlist(ling[,c("family1","family2")]))
length(uniqueFamilies)

```

```
## [1] 10
```

```

# Number of unique areas:
uniqueAreas = unique(unlist(ling[,c("area1","area2")]))
length(uniqueAreas)

```

```
## [1] 6
```

Cross-over between language families and areas:

```

tx = data.frame(lang= c(ling$l1,ling$l2),
                fam = c(ling$family1,ling$family2),
                area= c(ling$area1,ling$area2))
tx = tx[!duplicated(tx),]
table(tx$fam,tx$area)

```

##							
##		Europe	Greater	Mesopotamia	Indic	Inner	Asia N Coast Asia
##	Abkhaz-Adyge	0		1	0		0
##	Afro-Asiatic	0		1	0		0
##	Dravidian	0		0	3		0
##	Indo-European	11		2	1		0
##	Japonic	0		0	0		1
##	Koreanic	0		0	0		1
##	Mongolic	0		0	0		0
##	Sino-Tibetan	0		0	0		0
##	Turkic	0		1	0		0
##	Uralic	1		0	0		0
##							
##		Southeast	Asia				
##	Abkhaz-Adyge		0				
##	Afro-Asiatic		0				
##	Dravidian		0				
##	Indo-European		0				
##	Japonic		0				
##	Koreanic		0				
##	Mongolic		0				
##	Sino-Tibetan		1				
##	Turkic		0				
##	Uralic		0				

LMER models

Mixed effects model, predicting Linguistic similarity from cultural distances, with random intercept for family and area and random slope for cultural distance for family and area.

We compare a null model to a model with a fixed effect for cultural distance, with random intercepts for family and area, and random slopes for cultural distance by both.

```
m0 = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group) +
    (1 + cult.dist.center | area.group),
  data = ling
)
m1 = lmer(
  rho.center ~ 1 +
    cult.dist.center +
    (1 + cult.dist.center | family.group) +
    (1 + cult.dist.center | area.group),
  data = ling
)
anova(m0,m1)

## refitting model(s) with ML (instead of REML)

## Data: ling
## Models:
## m0: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 +
## m0:      cult.dist.center | area.group)
## m1: rho.center ~ 1 + cult.dist.center + (1 + cult.dist.center | family.group) +
## m1:      (1 + cult.dist.center | area.group)
##      Df      AIC      BIC logLik deviance Chisq Chi Df Pr(>Chisq)
## m0  8 1658.0 1694.7 -820.97  1642.0
## m1  9 1641.8 1683.2 -811.93  1623.8 18.091      1 2.106e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Cultural distance is significantly correlated with Linguistic similarity. Here are the model estimates:

```
summary(m1)

## Linear mixed model fit by REML ['lmerMod']
## Formula:
## rho.center ~ 1 + cult.dist.center + (1 + cult.dist.center | family.group) +
##      (1 + cult.dist.center | area.group)
##      Data: ling
##
## REML criterion at convergence: 1630.4
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.6291 -0.6811  0.0713  0.6658  4.6028
##
## Random effects:
##      Groups      Name              Variance Std.Dev.  Corr
## family.group (Intercept)    3.964e-01 6.296e-01
##      cult.dist.center 2.993e-02 1.730e-01 0.03
```

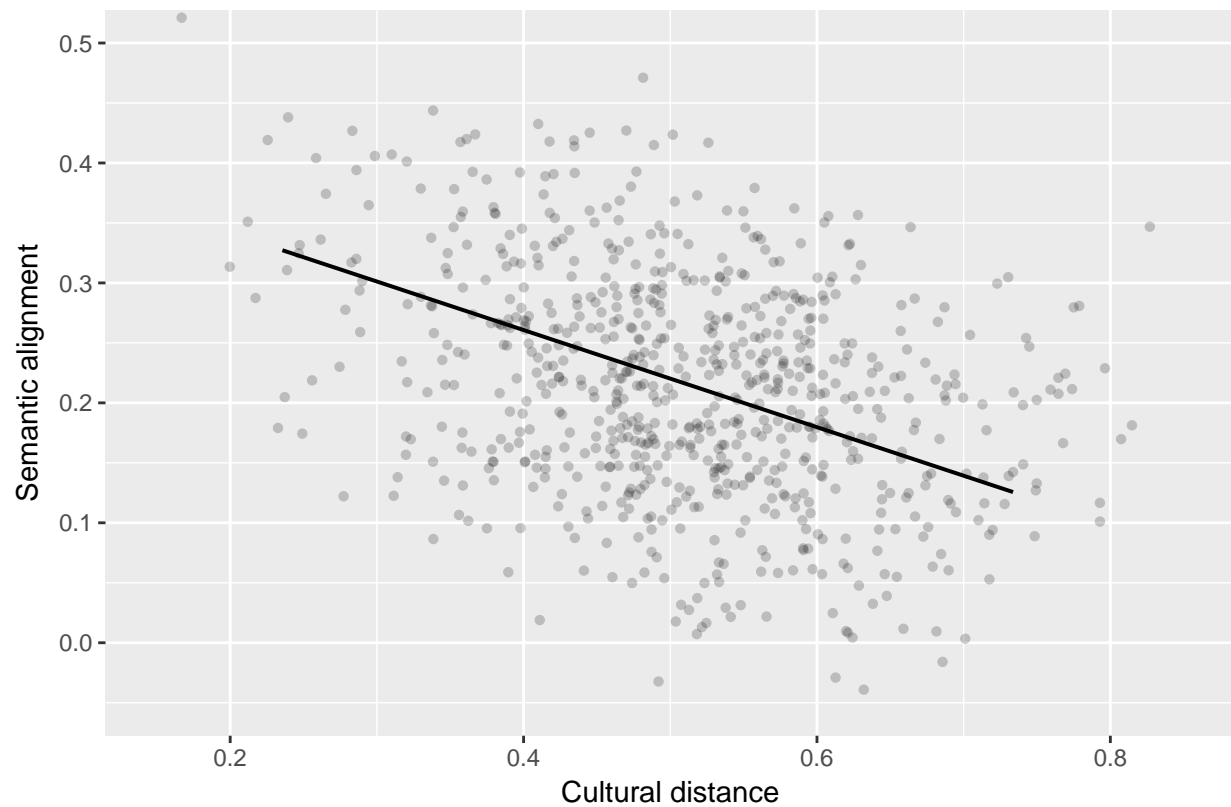
```
## area.group (Intercept) 0.000e+00 0.000e+00
## cult.dist.center 9.586e-15 9.791e-08 NaN
## Residual 4.697e-01 6.853e-01
## Number of obs: 733, groups: family.group, 48; area.group, 20
##
## Fixed effects:
## Estimate Std. Error t value
## (Intercept) -0.60832 0.10834 -5.615
## cult.dist.center -0.30894 0.05779 -5.346
##
## Correlation of Fixed Effects:
## (Intr)
## clt.dst.cnt -0.142
```

Plot the estimates, rescaling the variables back to the original units:

```
gx = sjp.lmer(m1, 'pred', 'cult.dist.center',
             prnt.plot = F)
```

```
## Warning: package 'bindrcpp' was built under R version 3.3.2
```

```
gx$plot$data$y = gx$plot$data$y *
  attr(ling$rho.center, "scaled:scale") +
  attr(ling$rho.center, "scaled:center")
gx$plot$data$res.p.y = gx$plot$data$res.p.y *
  attr(ling$rho.center, "scaled:scale") +
  attr(ling$rho.center, "scaled:center")
gx$plot$data$x = gx$plot$data$x *
  cdc.s + cdc.c
gx = gx$plot + coord_cartesian(ylim=c(-0.05,0.5),
                               xlim=c(0.15,0.85)) +
  xlab("Cultural distance") +
  ylab("Semantic alignment") +
  ggtitle("")
gx
```



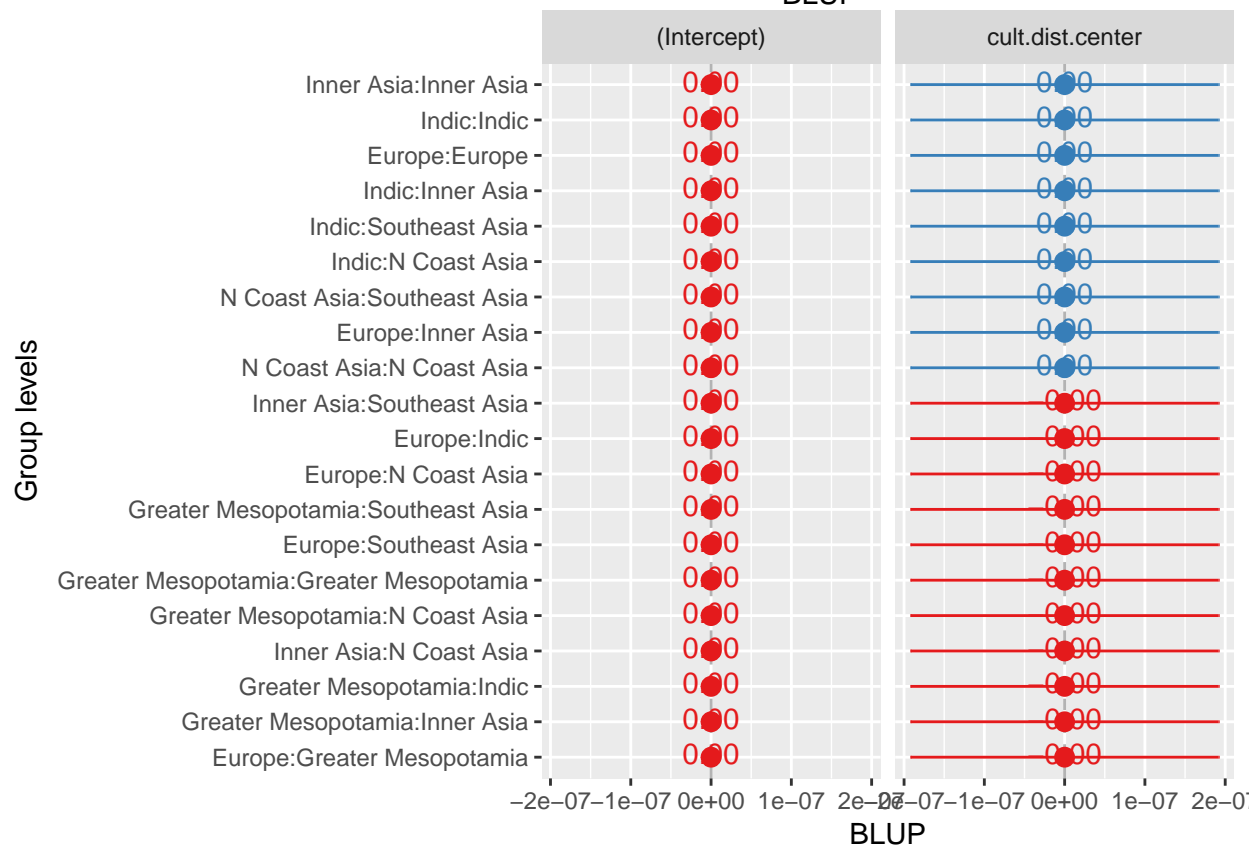
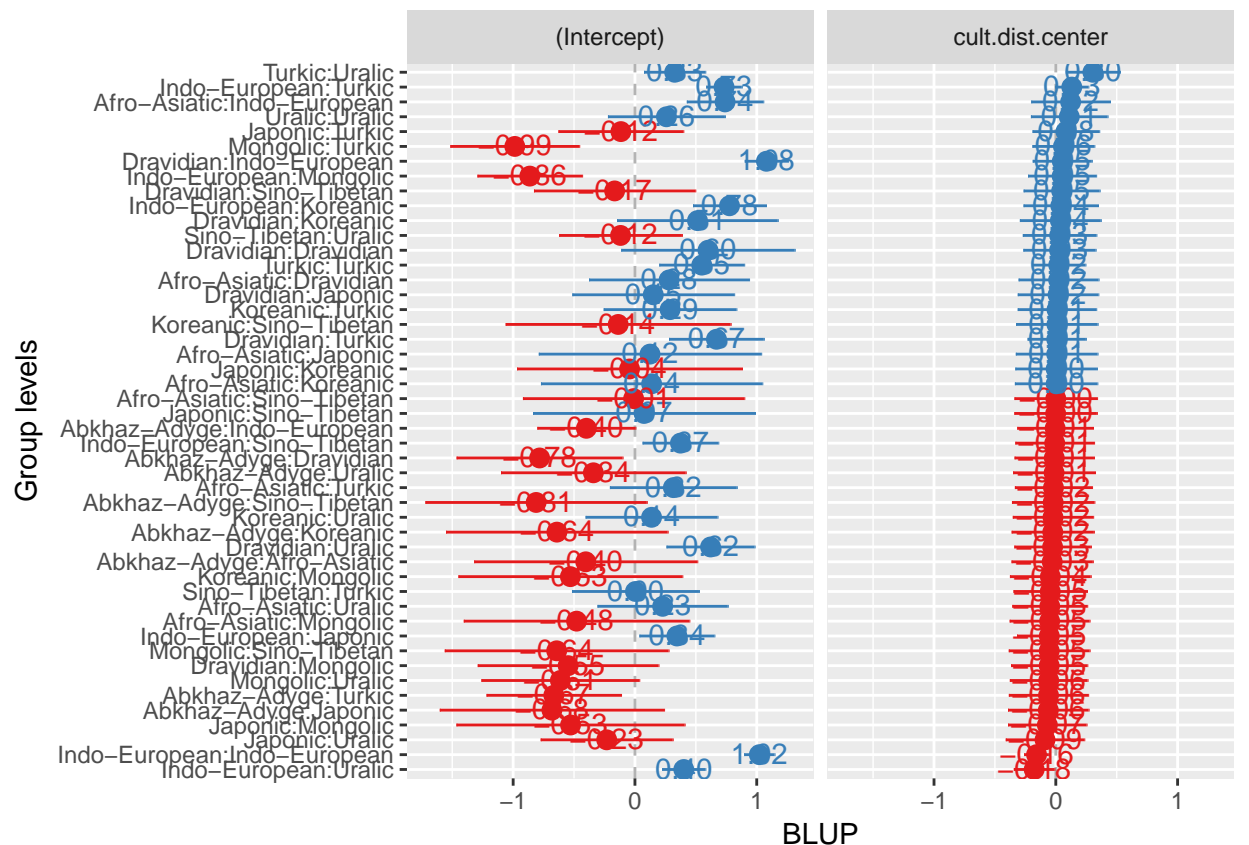
```
pdf(paste0("../results/stats/",datasetName,"/CulturalDistance_Rho_Graph.pdf"),
    height=2.5, width=2.5)
gx
dev.off()
```

```
## pdf
## 2
```

Plot the random effects:

```
sjp.lmer(m1,'re', sort.est = "cult.dist.center")
```

```
## Plotting random effects...
## Plotting random effects...
```

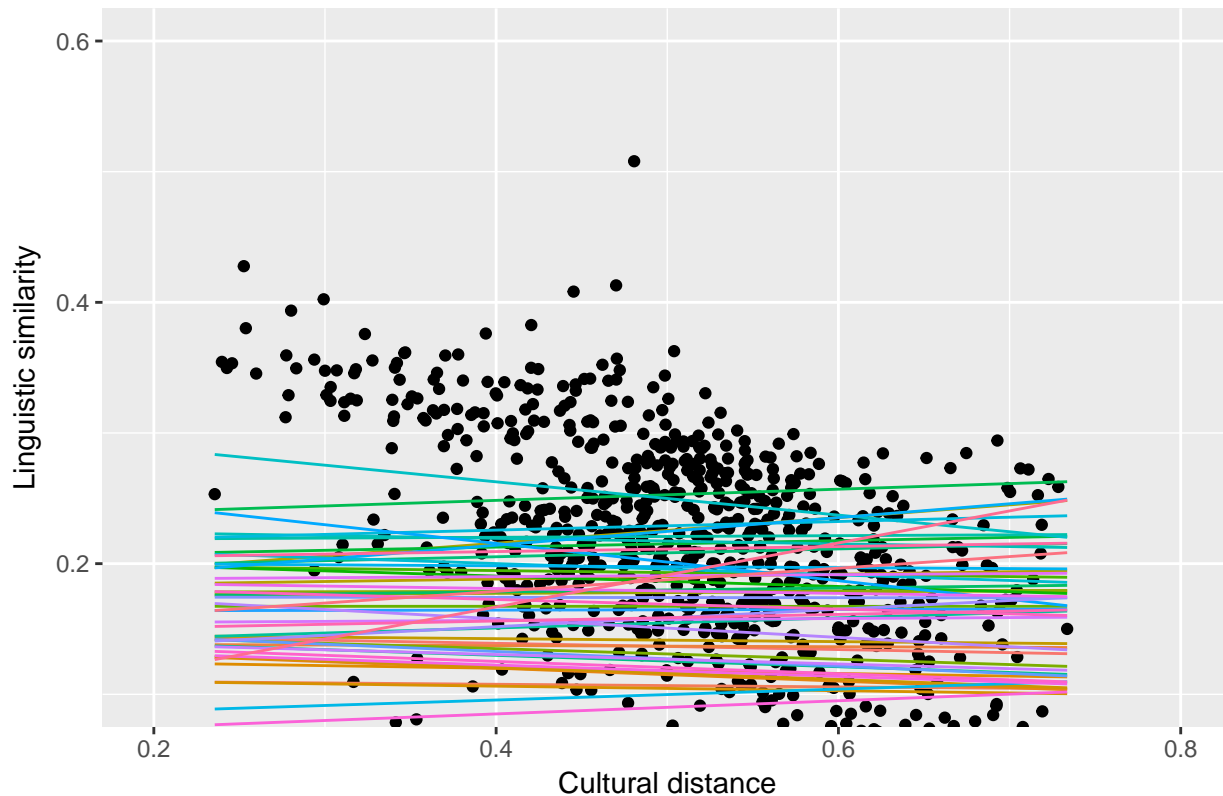



```

px = sjp.lmer(m1, 'rs.ri', prnt.plot = F)
dx = px$plot[[1]]$data
dx$x = dx$x * cdc.s + cdc.c
dx$y = dx$y *
  attr(ling$rho.center, "scaled:scale") +
  attr(ling$rho.center, "scaled:center")
ggplot(dx, aes(x, y)) +
  geom_point(data=ling,
             mapping=aes(x=as.numeric(cult.dist),
                         y=as.numeric(local_alignment))) +
  geom_line(mapping = aes(colour=grp)) +
  xlab("Cultural distance")+
  ylab("Linguistic similarity") +
  ggtitle("Language family pair random effects") +
  coord_cartesian(ylim=c(0.1,0.6),
                 xlim=c(0.2,0.8)) +
  theme(legend.position = "none")

```

Language family pair random effects



```

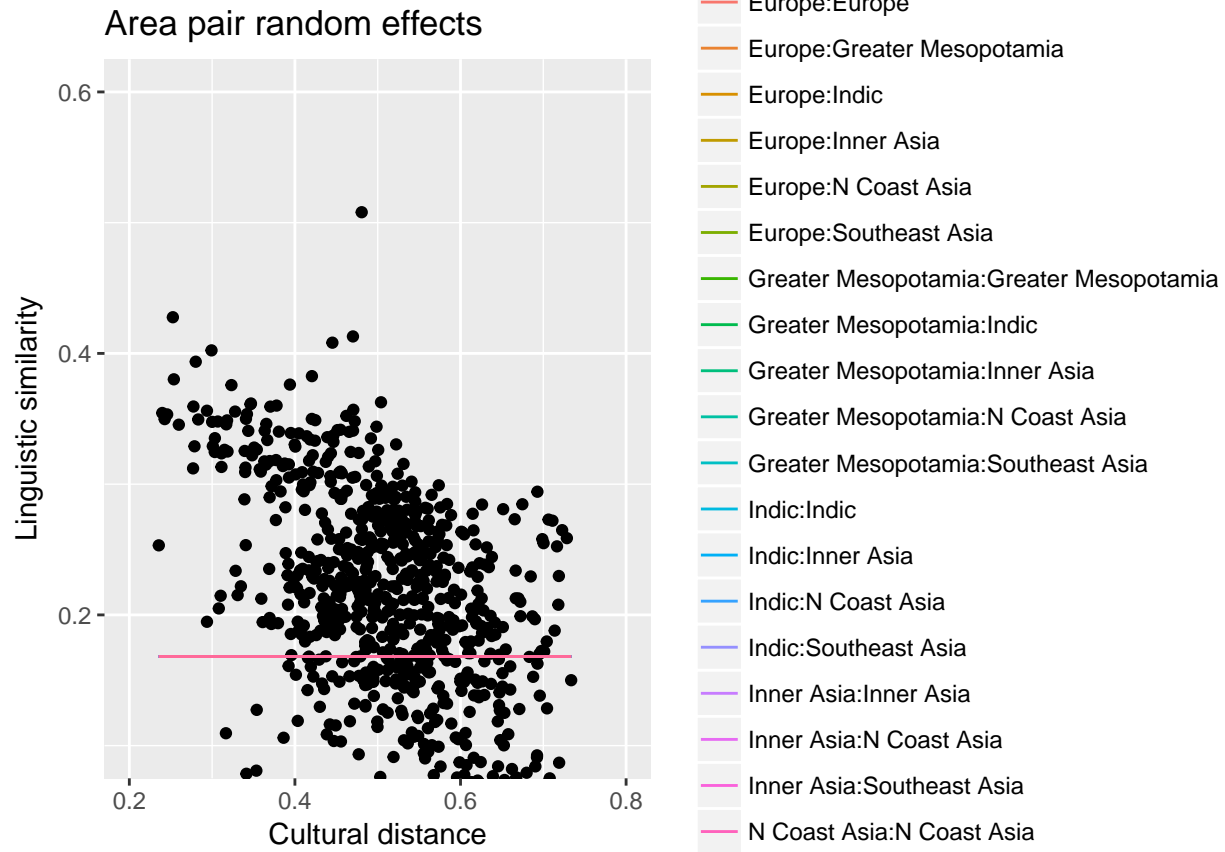
dx = px$plot[[2]]$data
dx$x = dx$x * cdc.s + cdc.c
dx$y = dx$y *
  attr(ling$rho.center, "scaled:scale") +
  attr(ling$rho.center, "scaled:center")
ggplot(dx, aes(x, y)) +
  geom_point(data=ling,
             mapping=aes(x=as.numeric(cult.dist),

```

```

y=as.numeric(local_alignment))) +
geom_line(mapping = aes(colour=grp)) +
xlab("Cultural distance")+
ylab("Linguistic similarity") +
ggtitle("Area pair random effects") +
coord_cartesian(ylim=c(0.1,0.6),
                xlim=c(0.2,0.8))

```



Note that the random slopes for area are set to 0. We can check whether taking language area out makes a difference:

```

m0b = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group),
  data = ling
)
m1b = lmer(
  rho.center ~ 1 +
    cult.dist.center +
    (1 + cult.dist.center | family.group),
  data = ling
)
anova(m0b,m1b)

```

```
## refitting model(s) with ML (instead of REML)
```

```
## Data: ling
```

```
## Models:
## m0b: rho.center ~ 1 + (1 + cult.dist.center | family.group)
## m1b: rho.center ~ 1 + cult.dist.center + (1 + cult.dist.center | family.group)
##      Df      AIC      BIC logLik deviance Chisq Chi Df Pr(>Chisq)
## m0b  5 1652.0 1674.9 -820.97  1642.0
## m1b  6 1635.8 1663.4 -811.93  1623.8 18.091      1 2.106e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(m1b)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula:
## rho.center ~ 1 + cult.dist.center + (1 + cult.dist.center | family.group)
##      Data: ling
##
## REML criterion at convergence: 1630.4
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.6291 -0.6811  0.0713  0.6658  4.6028
##
## Random effects:
##      Groups      Name      Variance Std.Dev. Corr
## family.group (Intercept)  0.39642  0.6296
##               cult.dist.center 0.02993  0.1730  0.03
## Residual                0.46969  0.6853
## Number of obs: 733, groups: family.group, 48
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)   -0.60832   0.10834  -5.615
## cult.dist.center -0.30894   0.05779  -5.346
##
## Correlation of Fixed Effects:
##              (Intr)
## clt.dst.cnt -0.142
```

The model is numerically almost exactly the same.

Tests within domains

Load data

Load distances for specific domains and match up to language family and area:

```
ling.dom = read.csv(
  lingDistancesByDomainFile,
  stringsAsFactors = F)

ling.dom = ling.dom[!is.na(ling.dom$cult.dist),]

ling.dom = ling.dom[(ling.dom$l1 %in% cultisos) &
  (ling.dom$l2 %in% cultisos),]
```

There are very few possible comparisons for Slovenian and Northern Sami, so we'll remove these:

```
ling.dom = ling.dom[!(ling.dom$l1=="se" || ling.dom$l2 == "se"),]
ling.dom = ling.dom[!(ling.dom$l1=="sl" || ling.dom$l2 == "sl"),]
```

Match family and area data:

```
ling.dom$family1 = 1[match(ling.dom$l1, 1$iso2),]$family
ling.dom$family2 = 1[match(ling.dom$l2, 1$iso2),]$family
ling.dom$area1 = 1[match(ling.dom$l1, 1$iso2),]$autotyp.area
ling.dom$area2 = 1[match(ling.dom$l2, 1$iso2),]$autotyp.area

# Paste language family names together,
# but order shouldn't matter, so sort first
fgroup = cbind(ling.dom$family1, ling.dom$family2)
fgroup = apply(fgroup, 1, sort)
ling.dom$family.group = apply(fgroup, 2, paste, collapse=":")

agroup = cbind(ling.dom$area1, ling.dom$area2)
agroup = apply(agroup, 1, sort)
ling.dom$area.group = apply(agroup, 2, paste, collapse=": ")
```

Center the data:

```
ling.dom$cult.dist.center = scale(ling.dom$cult.dist)
ling.dom$rho.center = scale(ling.dom$local_alignment)
```

LMER models

Test whether random slopes are warranted for family:

```
mD0 = lmer(
  rho.center ~ 1 +
    (1 | family.group) +
    (1 | area.group) +
    (1 | imputed_semantic_domain),
  data = ling.dom)
mD1 = lmer(
  rho.center ~ 1 +
    (1 | family.group) +
    (0 + cult.dist.center | family.group) +
    (1 | area.group) +
```

```

    (1 | imputed_semantic_domain),
    data = ling.dom)
mD2 = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group) +
    (1 | area.group) +
    (1 | imputed_semantic_domain),
    data = ling.dom)
anova(mD0,mD1,mD2)

## refitting model(s) with ML (instead of REML)

## Data: ling.dom
## Models:
## mD0: rho.center ~ 1 + (1 | family.group) + (1 | area.group) + (1 |
## mD0:      imputed_semantic_domain)
## mD1: rho.center ~ 1 + (1 | family.group) + (0 + cult.dist.center |
## mD1:      family.group) + (1 | area.group) + (1 | imputed_semantic_domain)
## mD2: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 |
## mD2:      area.group) + (1 | imputed_semantic_domain)
##      Df   AIC   BIC logLik deviance  Chisq Chi Df Pr(>Chisq)
## mD0  5 12830 12863 -6409.8    12820
## mD1  6 12592 12632 -6289.7    12580 240.163      1    < 2e-16 ***
## mD2  7 12588 12635 -6287.1    12574   5.322      1    0.02106 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Random slopes (and the correlation coefficient) for family improves the fit of the model.

Test the same for area:

```

mD3 = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group) +
    (1 | area.group) +
    (0 + cult.dist.center | area.group) +
    (1 | imputed_semantic_domain),
    data = ling.dom)
mD4 = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group) +
    (1 + cult.dist.center | area.group) +
    (1 | imputed_semantic_domain),
    data = ling.dom)
anova(mD2,mD3,mD4)

## refitting model(s) with ML (instead of REML)

## Data: ling.dom
## Models:
## mD2: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 |
## mD2:      area.group) + (1 | imputed_semantic_domain)
## mD3: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 |
## mD3:      area.group) + (0 + cult.dist.center | area.group) + (1 |
## mD3:      imputed_semantic_domain)
## mD4: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 +

```

```
## mD4:      cult.dist.center | area.group) + (1 | imputed_semantic_domain)
##      Df   AIC   BIC logLik deviance  Chisq Chi Df Pr(>Chisq)
## mD2  7 12588 12635 -6287.1    12574
## mD3  8 12572 12625 -6277.9    12556 18.4041      1 1.787e-05 ***
## mD4  9 12569 12629 -6275.3    12551  5.1295      1  0.02352 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Random slopes for area improve the fit of the model.

Test random slopes for domain:

```
mdom1 = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group) +
    (1 + cult.dist.center | area.group) +
    (1 | imputed_semantic_domain),
  data = ling.dom)
mdom2 = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group) +
    (1 + cult.dist.center | area.group) +
    (1 + cult.dist.center | imputed_semantic_domain),
  data = ling.dom)
anova(mdom1,mdom2)
```

refitting model(s) with ML (instead of REML)

```
## Data: ling.dom
## Models:
## mdom1: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 +
## mdom1:      cult.dist.center | area.group) + (1 | imputed_semantic_domain)
## mdom2: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 +
## mdom2:      cult.dist.center | area.group) + (1 + cult.dist.center |
## mdom2:      imputed_semantic_domain)
##      Df   AIC   BIC logLik deviance  Chisq Chi Df Pr(>Chisq)
## mdom1  9 12569 12629 -6275.3    12551
## mdom2 11 12527 12600 -6252.4    12505 45.851      2 1.105e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Random slope for domains significantly improves model.

Now we test the main effect of cultural distance:

```
mD5 = lmer(
  rho.center ~ 1 +
    (1 + cult.dist.center | family.group) +
    (1 + cult.dist.center | area.group) +
    (1 + cult.dist.center | imputed_semantic_domain),
  data = ling.dom)
mD6 = update(mD5, ~.+cult.dist.center)
anova(mD5,mD6)
```

refitting model(s) with ML (instead of REML)

```
## Data: ling.dom
## Models:
```



```
## mD5: rho.center ~ 1 + (1 + cult.dist.center | family.group) + (1 +
## mD5:      cult.dist.center | area.group) + (1 + cult.dist.center |
## mD5:      imputed_semantic_domain)
## mD6: rho.center ~ (1 + cult.dist.center | family.group) + (1 + cult.dist.center |
## mD6:      area.group) + (1 + cult.dist.center | imputed_semantic_domain) +
## mD6:      cult.dist.center
##      Df    AIC    BIC  logLik deviance  Chisq Chi Df Pr(>Chisq)
## mD5 11 12527 12600 -6252.4    12505
## mD6 12 12529 12609 -6252.3    12505 0.1742     1    0.6764
```

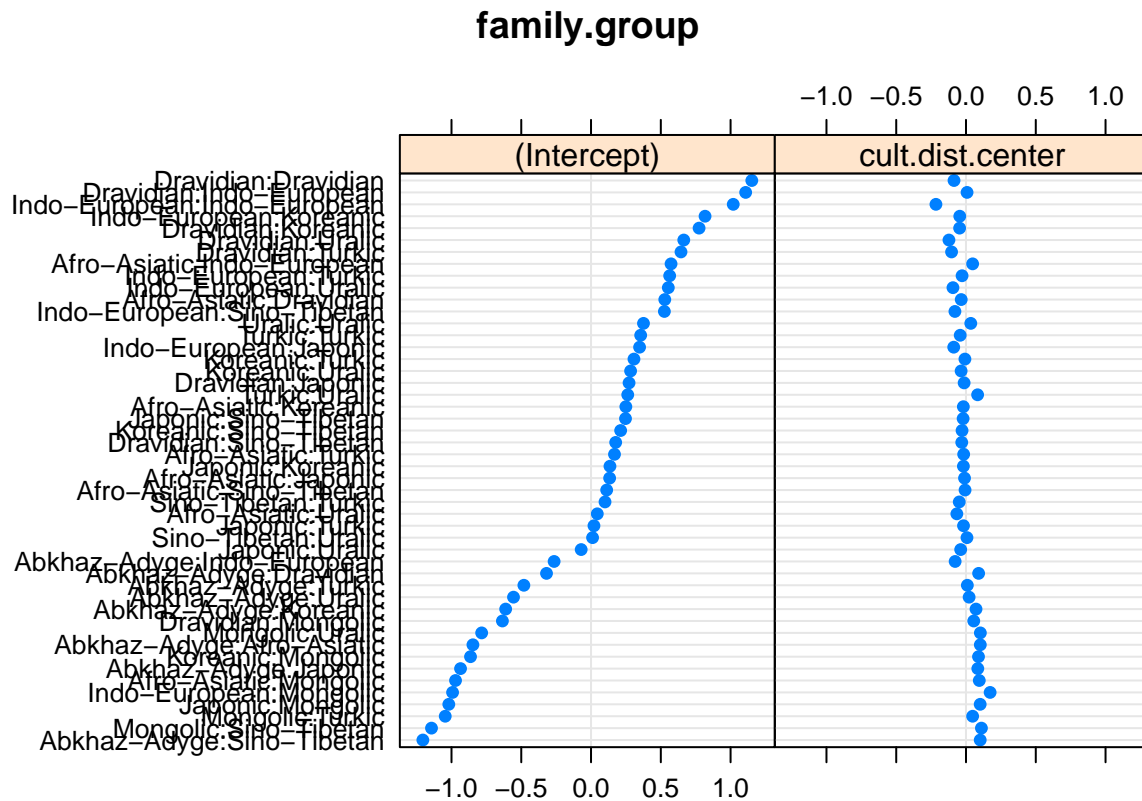
Summary of the final model, with random effects plot:

```
summary(mD6)
```

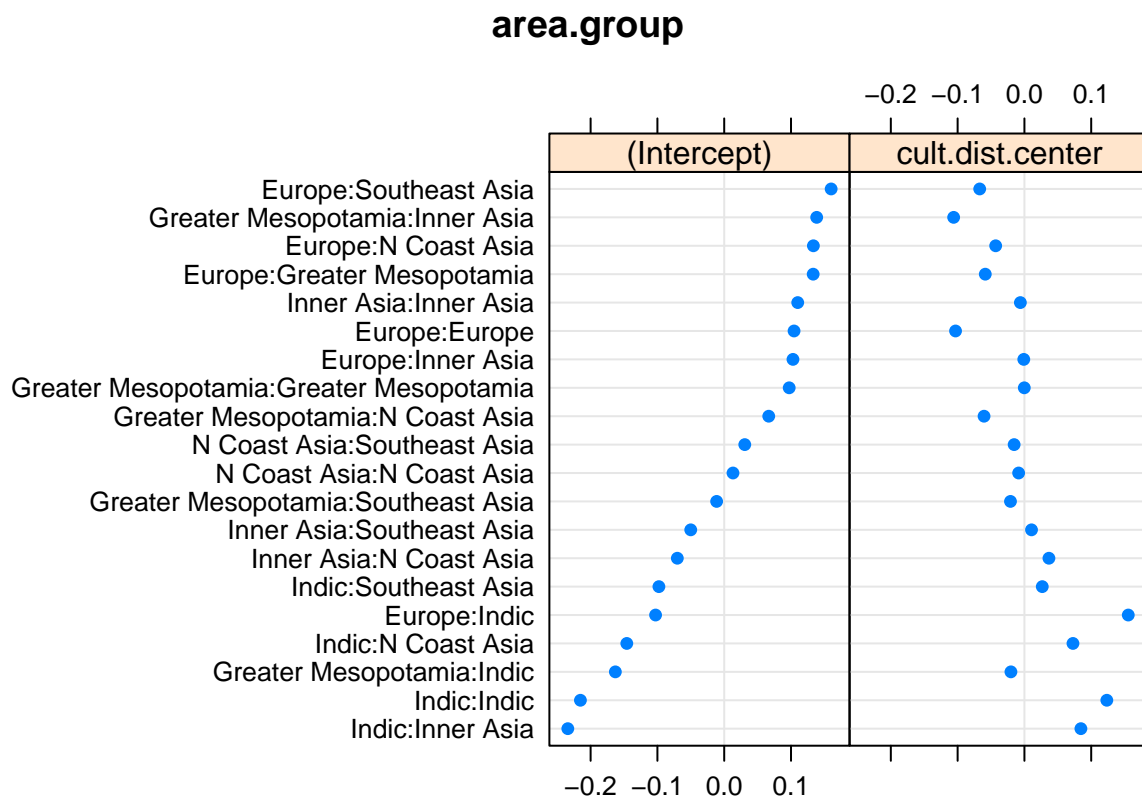
```
## Linear mixed model fit by REML ['lmerMod']
## Formula:
## rho.center ~ (1 + cult.dist.center | family.group) + (1 + cult.dist.center |
##      area.group) + (1 + cult.dist.center | imputed_semantic_domain) +
##      cult.dist.center
##      Data: ling.dom
##
## REML criterion at convergence: 12510.8
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -5.0342 -0.6137  0.0268  0.6099  3.8371
##
## Random effects:
##      Groups                Name                Variance Std.Dev. Corr
## family.group              (Intercept)          0.444533 0.66673
##                          cult.dist.center 0.011070 0.10521 -0.60
## area.group                (Intercept)          0.024293 0.15586
##                          cult.dist.center 0.008002 0.08945 -0.67
## imputed_semantic_domain (Intercept)          0.241060 0.49098
##                          cult.dist.center 0.007321 0.08556 -0.71
## Residual                    0.470128 0.68566
## Number of obs: 5858, groups:
## family.group, 48; area.group, 20; imputed_semantic_domain, 8
##
## Fixed effects:
##              Estimate Std. Error t value
## (Intercept)   -0.63238    0.20317  -3.113
## cult.dist.center -0.01869    0.04463  -0.419
##
## Correlation of Fixed Effects:
##              (Intr)
## clt.dst.cnt -0.576
```

```
dotplot(ranef(mD6))
```

```
## $family.group
```

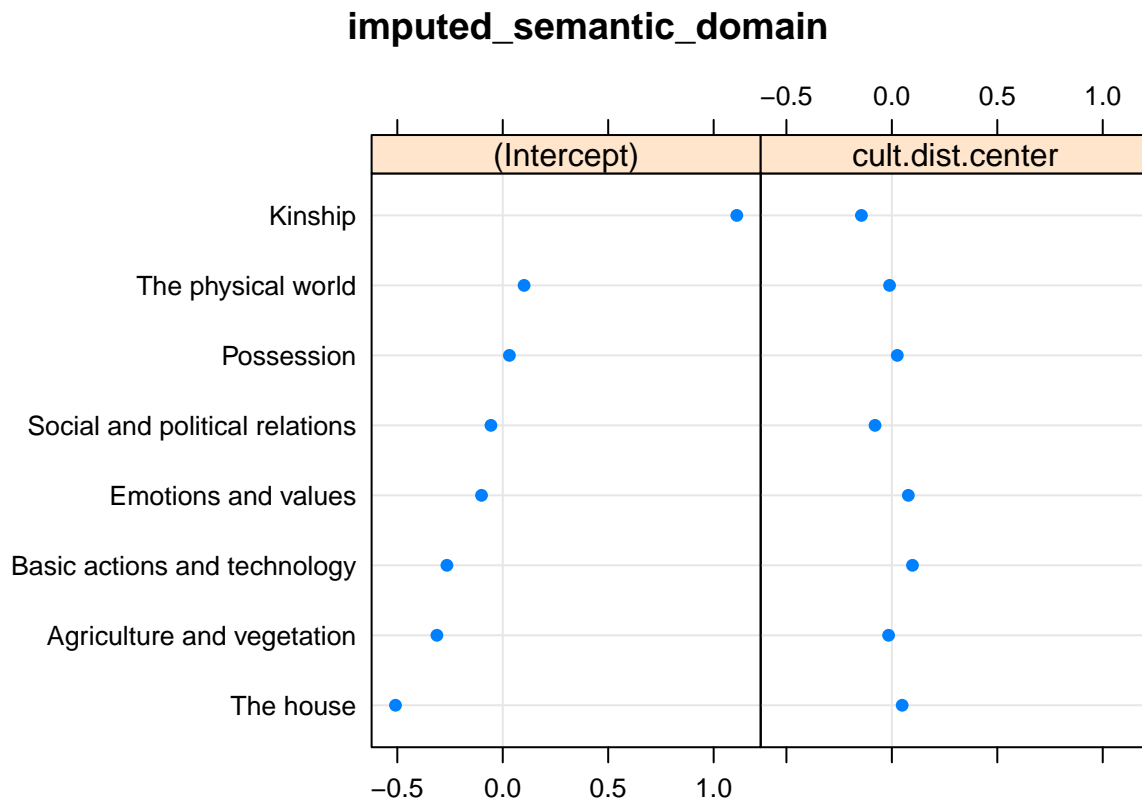


```
##
## $area.group
```



```
##
```

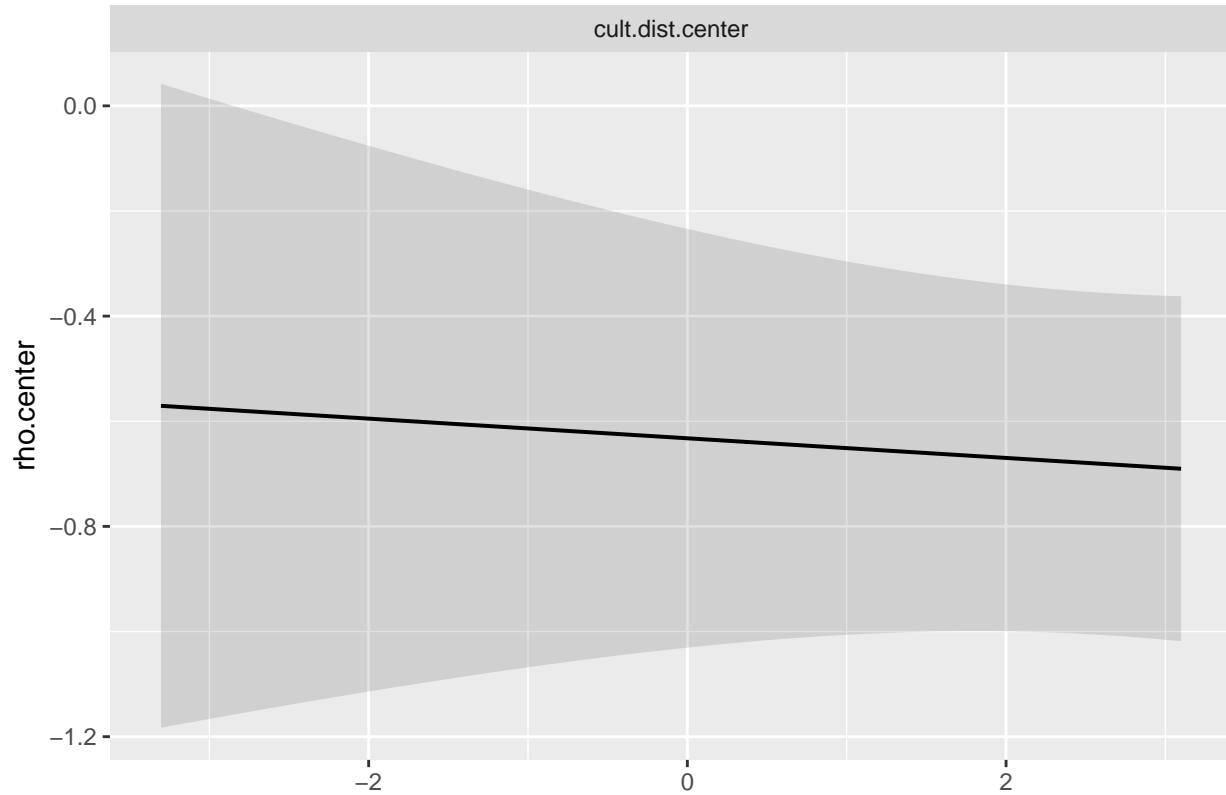
```
## $imputed_semantic_domain
```



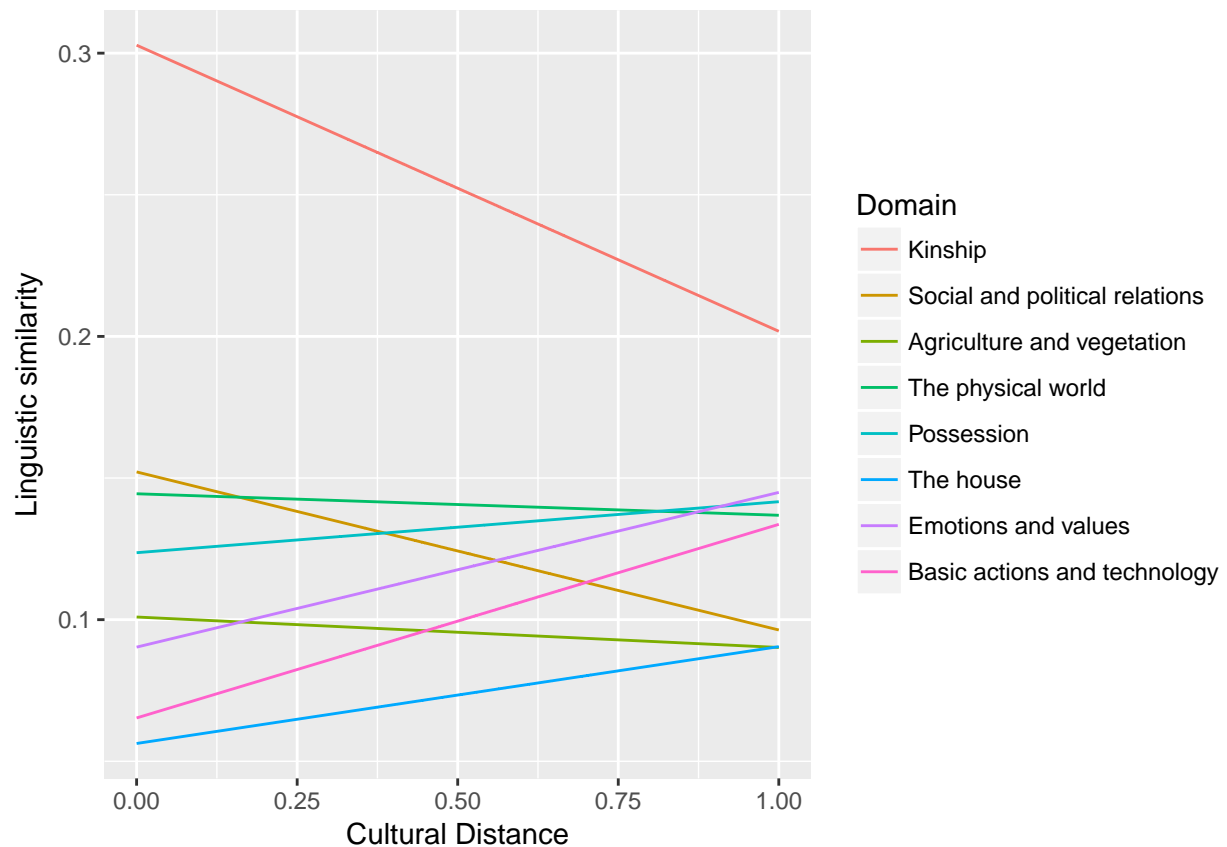
Plot the predicted relationships for each domain. The domains in the legend are sorted by the slope for cultural distance (greatest negative slope to greatest positive slope):

```
sjp.lmer(mD6,'eff', show.ci = T)
```

Marginal effects of model predictors



```
dom.order = ranef(mD6)$imputed_semantic_domain
dom.order = rownames(dom.order[order(dom.order$cult.dist.center),])
px = sjp.lmer(mD6,'rs.ri', show.ci = T, prnt.plot = F)
pdx = px$plot[[3]]$data
pdx$Domain = factor(pdx$grp, levels = dom.order)
pdx$x = pdx$x *
  attr(ling.dom$cult.dist.center,"scaled:scale") +
  attr(ling.dom$cult.dist.center,"scaled:center")
pdx$y = pdx$y *
  attr(ling.dom$rho.center,"scaled:scale") +
  attr(ling.dom$rho.center,"scaled:center")
ggplot(pdx,
  aes(x,y,colour=Domain)) +
  geom_line() +
  xlab("Cultural Distance") +
  ylab("Linguistic similarity")
```



Overall slope estimate for each domain:

```
ref.slopes.dom = data.frame(
  domain = rownames(ranef(mD6)$imputed_semantic_domain),
  slope = (fixef(mD6)["cult.dist.center"] +
    ranef(mD6)$imputed_semantic_domain[,2]))
ref.slopes.dom[order(ref.slopes.dom$slope),]
```

```
##           domain      slope
## 4           Kinship -0.163138433
## 6 Social and political relations -0.098486832
## 1   Agriculture and vegetation -0.034074828
## 8           The physical world -0.029551168
## 5           Possession  0.007048736
## 7           The house   0.030195874
## 3           Emotions and values  0.059418771
## 2   Basic actions and technology  0.079046059
```

Mantel tests

Read the historical distances for Indo-European, based on the phylogenetic distances.

Data prep

Load historical distances:

```
hist = read.csv("../data/trees/IndoEuropean_historical_distances.csv", stringsAsFactors = F)
hist = hist[!duplicated(hist[,1]),!duplicated(hist[,1])]
rownames(hist) = hist[,1]
hist = hist[,2:ncol(hist)]
hist.m = as.matrix(hist)
colnames(hist.m) = rownames(hist.m)
hist.m = hist.m/max(hist.m)
```

Read the cultural distances as a matrix:

```
cult.m = read.csv("../results/EA_distances/CulturalDistances.csv", stringsAsFactors = F)
rownames(cult.m) = cult.m[,1]
cult.m = cult.m[,2:ncol(cult.m)]
```

Convert the linguistic similarities to a matrix. This uses `igraph` to make an undirected graph from the long format with `local_alignment` as the edge weights, then output a matrix of adjacencies.

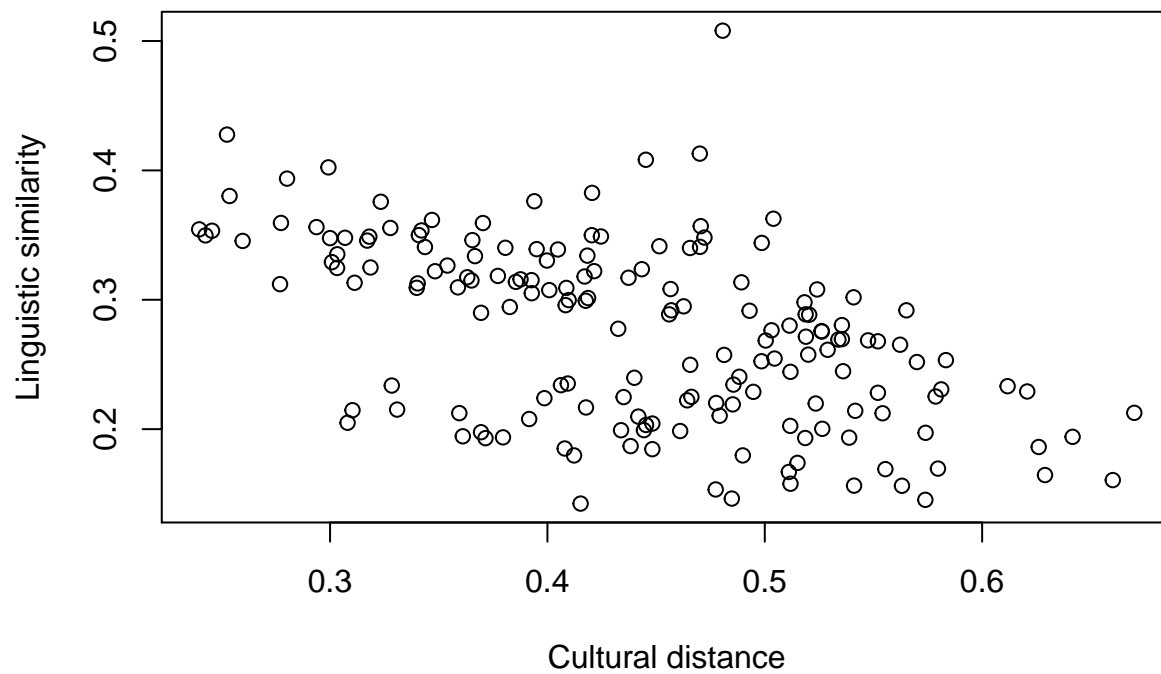
```
grph <- graph.data.frame(ling[,c("l1", "l2", "local_alignment")], directed=FALSE)
# add value as a weight attribute
ling.m = get.adjacency(grph, attr="local_alignment", sparse=FALSE)
rownames(ling.m) = 1[match(rownames(ling.m), l$iso2),]$Language2
colnames(ling.m) = 1[match(colnames(ling.m), l$iso2),]$Language2
```

Match the distance matrices

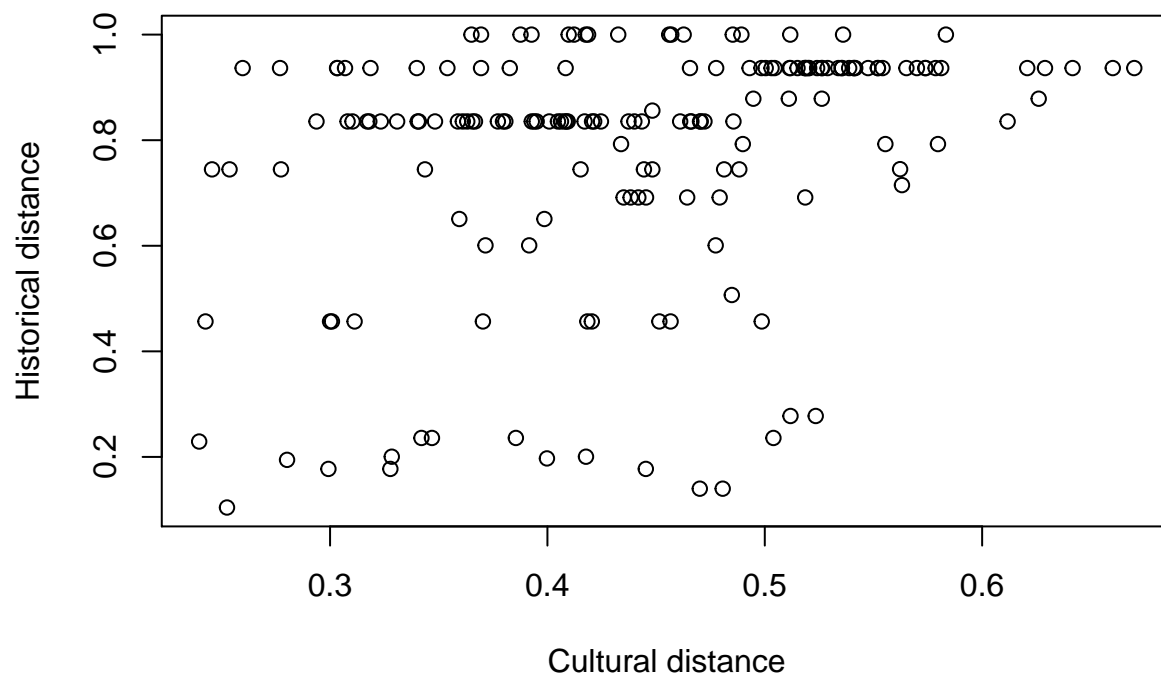
```
in.analysis = intersect(rownames(ling.m), rownames(cult.m))
in.analysis = intersect(in.analysis, rownames(hist.m))
cult.m2 = cult.m[in.analysis, in.analysis]
ling.m2 = ling.m[in.analysis, in.analysis]
hist.m2 = hist.m[in.analysis, in.analysis]
```

Note that there are only 19 languages with data on linguistic, cultural and historical distance.

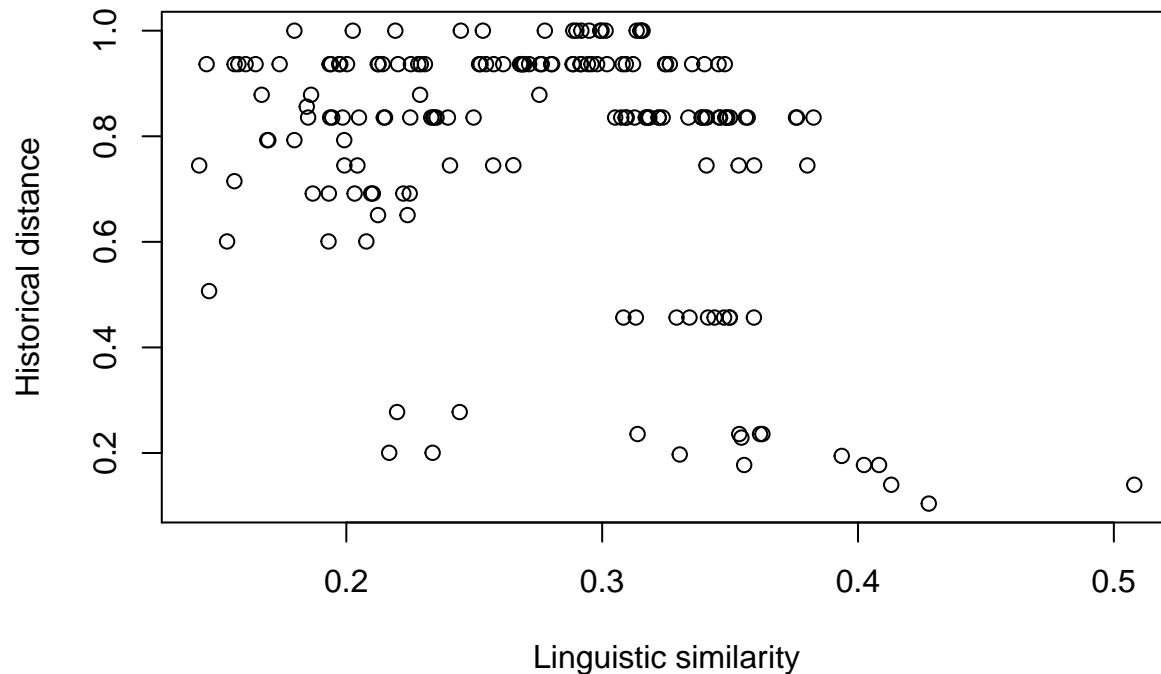
```
plot(as.dist(cult.m2), as.dist(ling.m2),
     xlab="Cultural distance",
     ylab="Linguistic similarity")
```



```
plot(as.dist(cult.m2),as.dist(hist.m2),
     xlab="Cultural distance",
     ylab="Historical distance")
```



```
plot(as.dist(ling.m2),as.dist(hist.m2),
     xlab="Linguistic similarity",
     ylab="Historical distance")
```



Tests

```
set.seed(1498)
```

Correlation between cultural and linguistic distances:

```
ecodist::mantel(as.dist(cult.m2) ~
  as.dist(ling.m2),
  nperm = 100000)
```

```
##      mantelr      pval1      pval2      pval3  llim.2.5% ulim.97.5%
## -0.5243289  0.9948700  0.0051400  0.0051700 -0.6278127 -0.3821098
```

Correlation between cultural and historical distances:

```
ecodist::mantel(as.dist(cult.m2) ~
  as.dist(hist.m2),
  nperm = 100000)
```

```
##      mantelr      pval1      pval2      pval3  llim.2.5% ulim.97.5%
##  0.3243830  0.0128500  0.9871600  0.0137800  0.2286808  0.4280939
```

Correlation between linguistic and historical distances:

```
ecodist::mantel(as.dist(ling.m2) ~
  as.dist(hist.m2),
  nperm = 100000)
```

```
##      mantelr      pval1      pval2      pval3  llim.2.5% ulim.97.5%
## -0.3372882  0.9863800  0.0136300  0.0163000 -0.5054090 -0.1742784
```

Run a mantel test comparing the Linguistic similaritys to the cultural distances, controlling for the historical distance between languages:


```
ecodist::mantel(as.dist(ling.m2)~
                as.dist(cult.m2) +
                as.dist(hist.m2),
                nperm = 100000)
```

```
##      mantelr      pval1      pval2      pval3  llim.2.5% ulim.97.5%
## -0.4659407  0.9896600  0.0103500  0.0111900 -0.6027438 -0.3108281
```

Run a mantel test comparing the linguistic similarities to the historical similarities, controlling for cultural distance:

```
ecodist::mantel(as.dist(ling.m2)~
                as.dist(hist.m2) +
                as.dist(cult.m2),
                nperm = 100000)
```

```
##      mantelr      pval1      pval2      pval3  llim.2.5%  ulim.97.5%
## -0.20758646  0.92203000  0.07798000  0.14086000 -0.41440025  0.04182923
```

Are the relevant D-place features most predictive?

Long format to line up comparisons:

```
ling.dom.wide = ling.dom[,c("l1", "l2",  
    'imputed_semantic_domain',  
    "local_alignment", "cult.dist")]  
ling.dom.wide = reshape(ling.dom.wide,  
    idvar = c("l1", "l2"),  
    timevar = "imputed_semantic_domain",  
    direction = "wide")  
ling.dom.wide = cbind(ling.dom.wide[,1:2],  
    ling.dom.wide[,3:ncol(ling.dom.wide)][  
    order(names(ling.dom.wide[,3:ncol(ling.dom.wide)]))  
    ])  
snames = c("Agri", "Tech", "Emot", "Kin", "Poss", "Soc", "Hous", "Wrld")  
names(ling.dom.wide) = c("l1", "l2",  
    paste0("L.", snames),  
    paste0("C.", snames))
```

Raw correlation between each pair of domains

```
compareAllDomains =  
    cor(ling.dom.wide[,  
        grepl("L\\. ", names(ling.dom.wide))],  
        ling.dom.wide[,  
            grepl("C\\. ", names(ling.dom.wide))],  
        use="complete.obs")  
  
round(compareAllDomains, 2)
```

```
##          C.Agri C.Tech C.Emot C.Kin C.Poss C.Soc C.Hous C.Wrld  
## L.Agri  -0.47  -0.39  -0.42  -0.41  -0.33  -0.48  -0.34  -0.52  
## L.Tech  -0.26  -0.23  -0.26  -0.34  -0.18  -0.27  -0.21  -0.33  
## L.Emot  -0.27  -0.19  -0.27  -0.42  -0.18  -0.26  -0.23  -0.36  
## L.Kin   -0.28  -0.18  -0.17  -0.27  -0.19  -0.27  -0.19  -0.28  
## L.Poss  -0.44  -0.39  -0.42  -0.52  -0.34  -0.48  -0.38  -0.53  
## L.Soc   -0.51  -0.35  -0.42  -0.51  -0.33  -0.49  -0.39  -0.53  
## L.Hous  -0.30  -0.16  -0.14  -0.21  -0.18  -0.23  -0.18  -0.24  
## L.Wrld  -0.18  -0.12  -0.08  -0.10  -0.09  -0.15  -0.04  -0.17
```

References

Bouckaert, Remco, Philippe Lemey, Michael Dunn, Simon J. Greenhill, Alexander V. Alekseyenko, Alexei J. Drummond, Russell D. Gray, Marc A. Suchard, and Quentin D. Atkinson (2012). Mapping the origins and expansion of the Indo-European language family. *Science*, 337(6097), 957-960.