

lambda vs w plot

February 19, 2019

0.1 Try to plot the lambda function vs the baseline covariates

Date: 2019-2-20

```
In [2]: setwd('/Users/yaolanqiu/Desktop/NYU/rotation/Rotation2/Week3/from dr.tarpey')
        source("cvxcluster-0513.R")
```

```
In [12]: library('lme4')
```

```
In [13]: # read in data
        dat = read.table("hcaf.dat", header=T)
        d0 = dat[dat$trt == 0,]
        d1 = dat[dat$trt == 1,]
        head(dat)
```

subj	trt	y	age	BaselineCGI	t1	responder
2497	0	25	29	4	0	0
2497	0	18	29	4	1	0
2497	0	11	29	4	2	0
2497	0	9	29	4	3	0
2497	0	19	29	4	4	0
2497	0	15	29	4	5	0

```
In [14]: # create new covariates
        cov01 = rnorm(length(unique(d0$subj)),5,1)
        cov02 = rnorm(length(unique(d0$subj)),10,1)
        newcov0 = data.frame(subj = unique(d0$subj), newcov1 = cov01, newcov2 = cov02)
        d0 = merge(d0,newcov0, by = 'subj')
        head(d0)
```

subj	trt	y	age	BaselineCGI	t1	responder	newcov1	newcov2
2497	0	25	29	4	0	0	5.115883	10.0736
2497	0	18	29	4	1	0	5.115883	10.0736
2497	0	11	29	4	2	0	5.115883	10.0736
2497	0	9	29	4	3	0	5.115883	10.0736
2497	0	19	29	4	4	0	5.115883	10.0736
2497	0	15	29	4	5	0	5.115883	10.0736

```
In [15]: # create new covariates
cov01 = rnorm(length(unique(d1$subj)),10,1)
cov02 = rnorm(length(unique(d1$subj)),5,1)
newcov1 = data.frame(subj = unique(d1$subj), newcov1 = cov01, newcov2 = cov02)
d1 = merge(d1,newcov1, by = 'subj')
head(d1)
```

subj	trt	y	age	BaselineCGI	t1	responder	newcov1	newcov2
2500	1	24	29	4	0	0	10.031118	4.965798
2500	1	22	29	4	1	0	10.031118	4.965798
2500	1	19	29	4	2	0	10.031118	4.965798
2500	1	25	29	4	3	0	10.031118	4.965798
2509	1	24	31	4	0	1	9.820069	7.081584
2509	1	10	31	4	1	1	9.820069	7.081584

```
In [16]: # new covariate, which is the combination of the two new covariates
# let's make it a simple summation first
d0$newcov = d0$newcov1 + d0$newcov2
d1$newcov = d1$newcov1 + d1$newcov2
```

```
In [18]: # Fit LME
fit_d0 = lmer(y ~ t1 + I(t1^2) + newcov + newcov * t1 + newcov * I(t1^2) + (t1+I(t1^2) | subj),
              data = d0, REML = FALSE)
fit_d1 = lmer(y ~ t1 + I(t1^2) + newcov + newcov * t1 + newcov * I(t1^2) + (t1+I(t1^2) | subj),
              data = d1, REML = FALSE)
```

singular fit

```
In [20]: summary(fit_d0)
```

Linear mixed model fit by maximum likelihood ['lmerMod']
Formula: y ~ t1 + I(t1^2) + newcov + newcov * t1 + newcov * I(t1^2) + (t1 + I(t1^2) | subj)
Data: d0

AIC	BIC	logLik	deviance	df.resid
5970.0	6033.6	-2972.0	5944.0	965

Scaled residuals:

Min	1Q	Median	3Q	Max
-4.2516	-0.5961	-0.0336	0.5374	2.9550

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
subj	(Intercept)	2.1349	1.4611	
	t1	5.1128	2.2612	0.96
	I(t1^2)	0.1213	0.3482	-0.98 -0.88
Residual		16.1672	4.0208	

Number of obs: 978, groups: subj, 162

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	29.31307	3.39179	8.642
t1	-9.08822	3.28602	-2.766
I(t1 ²)	0.98201	0.55240	1.778
newcov	-0.41718	0.22569	-1.849
t1:newcov	0.32262	0.21865	1.476
I(t1 ²):newcov	-0.03216	0.03675	-0.875

Correlation of Fixed Effects:

	(Intr)	t1	I(t1 ²)	newcov	t1:nwc
t1		-0.344			
I(t1 ²)	0.242	-0.918			
newcov	-0.996	0.343	-0.241		
t1:newcov	0.343	-0.996	0.914	-0.344	
I(t1 ²):nwc	-0.241	0.914	-0.996	0.242	-0.918

convergence code: 0
singular fit

In [21]: `summary(fit_d1)`

Linear mixed model fit by maximum likelihood ['lmerMod']

Formula: $y \sim t1 + I(t1^2) + newcov + newcov * t1 + newcov * I(t1^2) + (t1 + I(t1^2) | subj)$

Data: d1

AIC	BIC	logLik	deviance	df.resid
7281.4	7347.9	-3627.7	7255.4	1218

Scaled residuals:

Min	1Q	Median	3Q	Max
-3.2540	-0.5487	-0.0448	0.5174	3.4531

Random effects:

Groups	Name	Variance	Std.Dev.	Corr
subj	(Intercept)	8.0620	2.8394	
	t1	7.5757	2.7524	0.07
	I(t1 ²)	0.1552	0.3939	-0.22 -0.93
Residual		11.7507	3.4279	

Number of obs: 1231, groups: subj, 196

Fixed effects:

	Estimate	Std. Error	t value
(Intercept)	27.49879	3.27545	8.395

t1	-9.88284	2.90459	-3.402
I(t1^2)	1.16388	0.44727	2.602
newcov	-0.25199	0.21587	-1.167
t1:newcov	0.36967	0.19139	1.932
I(t1^2):newcov	-0.05253	0.02945	-1.784

Correlation of Fixed Effects:

	(Intr)	t1	I(t1^2)	newcov	t1:nwc
t1		-0.338			
I(t1^2)		0.210	-0.934		
newcov		-0.996	0.337	-0.209	
t1:newcov		0.337	-0.996	0.930	-0.338
I(t1^2):nwc		-0.209	0.931	-0.996	0.210 -0.934

0.1.1 Plot the $\lambda(x)$ v.s. w figure

The λ here is just

$$\lambda(z_i|w_i) = \frac{f_2(z_i|w_i)}{f_1(z_i|w_i) + f_2(z_i|w_i)}$$

Steps:

1. Fit LME, calculate the mean and variance of MVN for drug group and placebo group, separately.
2. Generate data from the above MVN by using the mean and variance calculated from the LME
3. Input values (the value from above generation) and selected w values into the λ function.
4. Draw the plot

```
In [23]: beta0 = as.matrix(fixef(fit_d0)[1:3])
gamma0 = as.matrix(fixef(fit_d0)[4:6])
beta1 = as.matrix(fixef(fit_d1)[1:3])
gamma1 = as.matrix(fixef(fit_d1)[4:6])
```

```
In [26]: lambda2 = function(xx, mu0, mu1, sigma0, sigma1){
```

```
  quadratic0 = function(a,b) {
    X = matrix(c(a,b),nrow=2)
    Q = (-1/2)*t(X-mu0)%*%solve(sigma0)%*%(X-mu0)
  }
  quadratic1 = function(a,b) {
    X = matrix(c(a,b),nrow=2)
    Q = (-1/2)*t(X-mu1)%*%solve(sigma1)%*%(X-mu1)
  }
```

```
  f0 = (1/(2*pi))*(1/sqrt(det(sigma0)))*exp(quadratic0(xx[1],xx[2]))
  f1 = (1/(2*pi))*(1/sqrt(det(sigma1)))*exp(quadratic1(xx[1],xx[2]))
```

```

    return(f1/(f1 + f0))
}

```

```

In [30]: drg_group = c()
pat_group = c()
Ww = c()
for(w in seq(10,20,1)){

  m0 = beta0 + gamma0 * w; m0 = m0[2:3]
  m1 = beta1 + gamma1 * w; m1 = m1[2:3]
  D0 = as.matrix(VarCorr(fit_d0)$subj)[2:3, 2:3]
  D1 = as.matrix(VarCorr(fit_d1)$subj)[2:3, 2:3]

  mu0 = matrix(m0, nrow=2)
  sigma0 = D0
  mu1 = matrix(m1, nrow=2)
  sigma1 = D1

  pat = mvrnorm(10, mu0, sigma0)
  drg = mvrnorm(10, mu1, sigma1)

  for(i in 1:10){
    pat_group = c(pat_group, lambda2(pat[i,], mu0, mu1, sigma0, sigma1))
  }

  for(i in 1:10){
    drg_group = c(drg_group, lambda2(drg[i,], mu0, mu1, sigma0, sigma1))
  }

  Ww = c(Ww, rep(w, 10))
}

```

```

In [35]: options(repr.plot.width=4, repr.plot.height=3)
plot(Ww, drg_group, cex = 0.5, col = 'red', ylab = 'lambda', xlab= 'w')
points(Ww+0.1, pat_group, cex = 0.5)

```

