

Hasher-Lind Normal

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The function that we will use as our test of Hasofer-Lind is

$$Z = x_1 - x_2 + 0.1 \cos(4x_1)$$

. Let's say that

$$X_1 \sim N(-0.2, 0.5)$$

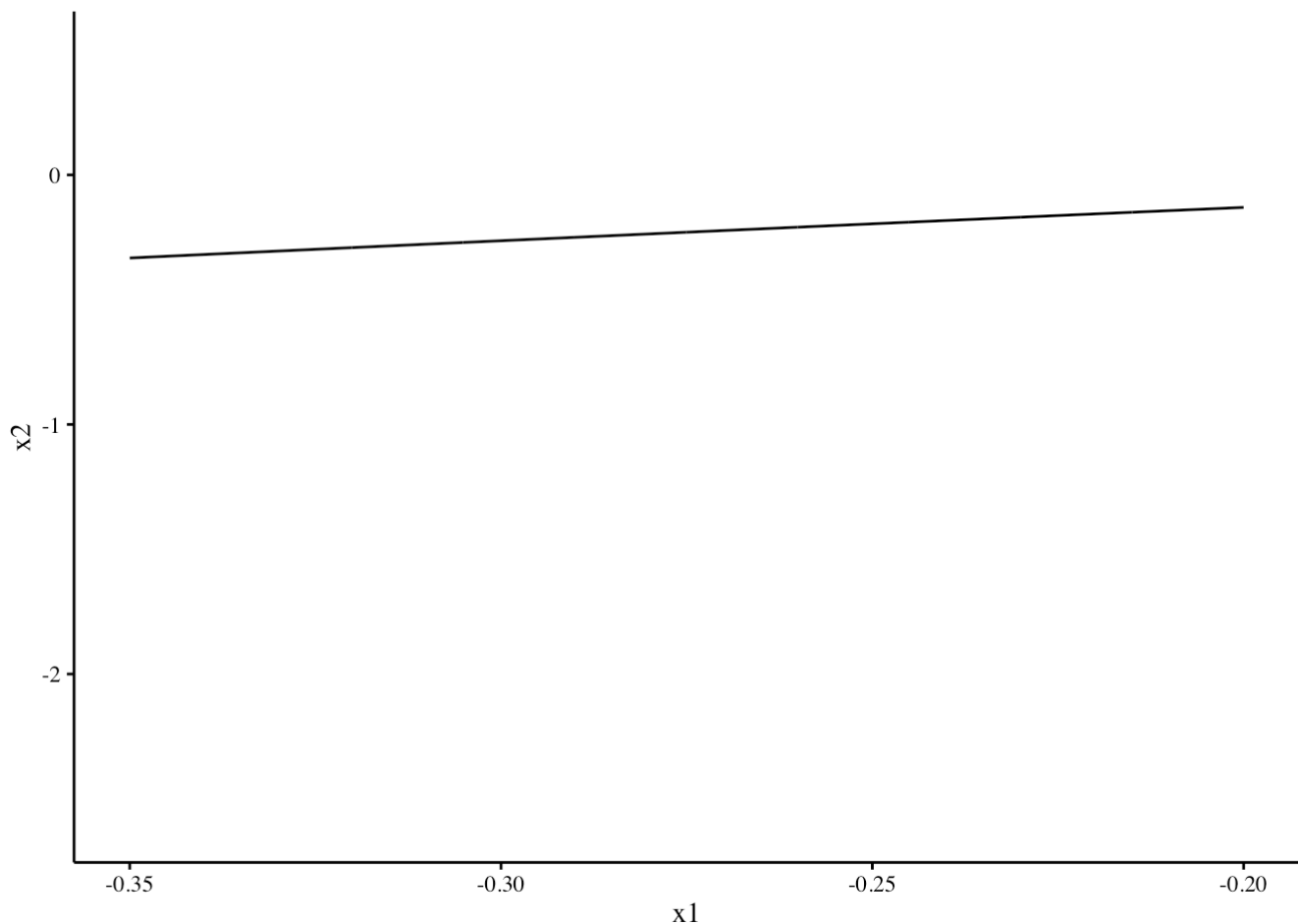
and

$$X_2 \sim N(-2.5, 2.5).$$

The failure surface for this function (i.e., where $Z = 0$) occurs at

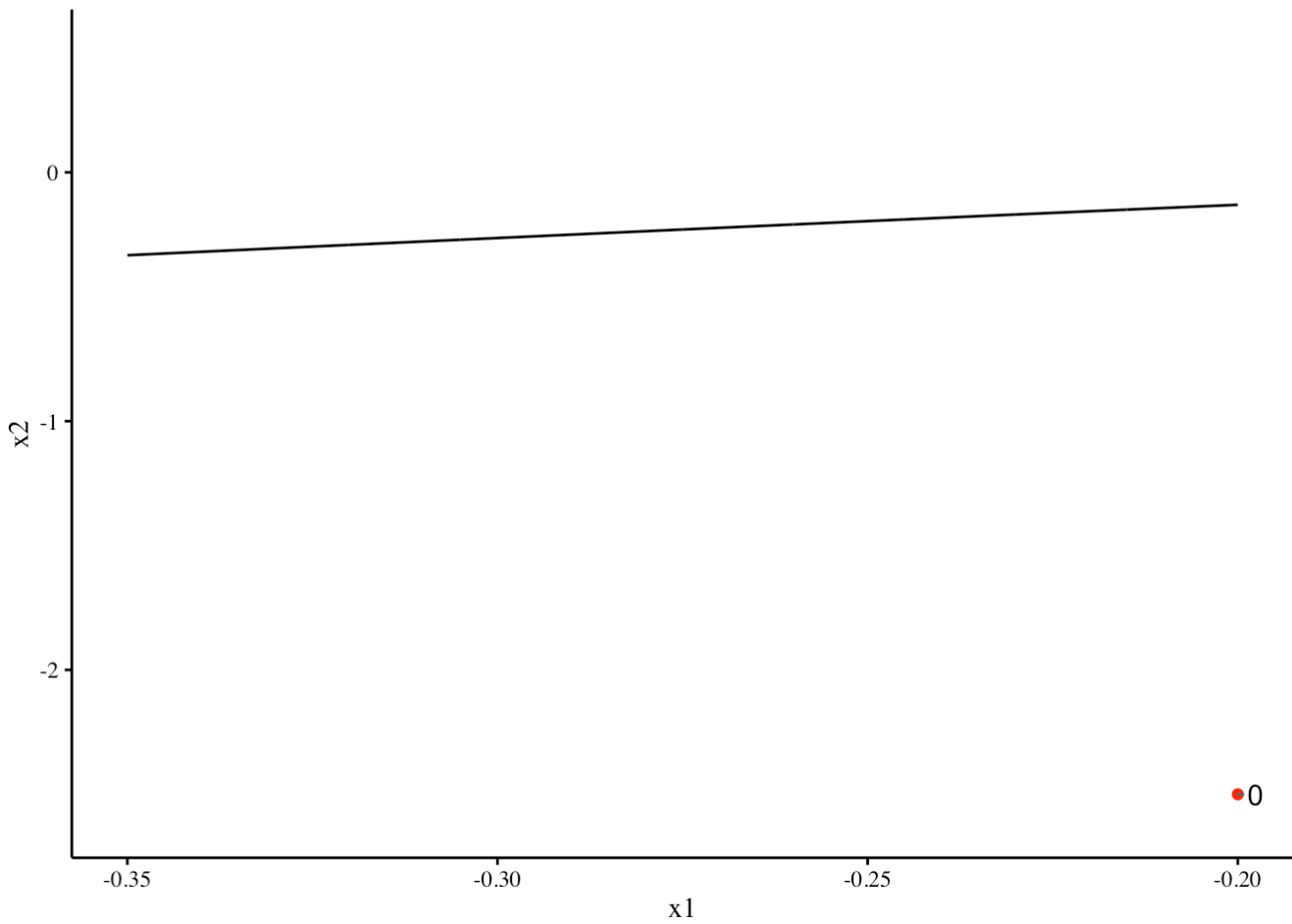
$$x_2 = x_1 + 0.1 \cos(4x_1)$$

.



The HL procedure begins at the mean point.

```
mux1 = -0.2  
mux2 = -2.5  
mainplot = mainplot + geom_point(data=data.frame(X1=mux1,X2=mux2), color="red") +  
  geom_text_repel(data=data.frame(X1=mux1,X2=mux2), label=0)  
print(mainplot)
```



Next we normalize the x_i and find a search direction.

```

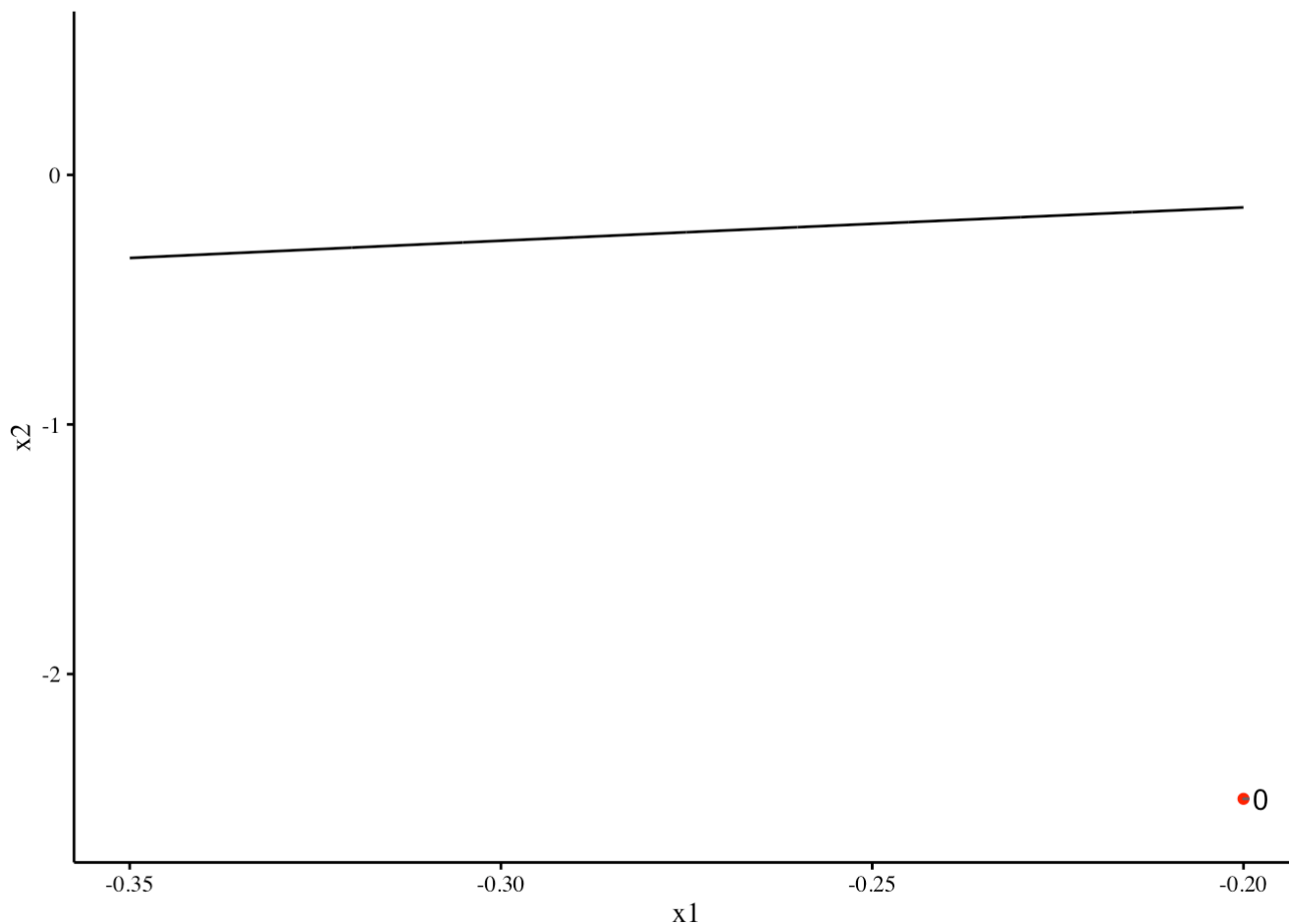
sigmax1 = 0.5
sigmax2 = 2.5
iteration1 = data.frame(X1 = mux1, X2 = mux2)
iteration1$X1prime = (iteration1$X1-mux1)/sigmax1
iteration1$X2prime = (iteration1$X2-mux2)/sigmax2

dg = c(dZdx1(iteration1), dZdx2(iteration1))
dgprime = dg*c(sigmax1, sigmax2)
absgsquard = sum(dgprime*dgprime)
xprime = 1/absgsquard*(sum(dgprime*c(iteration1$X1prime,iteration1$X2prime))
                      - Z(iteration1))*dgprime
iteration2 = data.frame(X1 = xprime[1]*sigmax1+mux1,X2 = xprime[2]*sigmax2+mux2 )

beta = sqrt(sum(xprime^2))
absg = abs(Z(iteration2))

#mainplot = mainplot + geom_point(data=iteration2, color="red") + geom_text_repel(data=iteration
2, label=1)
print(mainplot)

```



```

iteration2$lab = "1"
allits = iteration2

```

After one iteration $\beta = 0.9179493$, and $|Z| = 0.0060349$.

Now let's turn this into an procedure that runs until either $\Delta\beta$ or $|Z|$ is less than 10^{-6} .

```
delta = 1e-6
converged = 0
it = 2
beta_old = beta
old_iteration = iteration2
while (!(converged)){
  old_iteration$X1prime = (old_iteration$X1-mux1)/sigmax1
  old_iteration$X2prime = (old_iteration$X2-mux2)/sigmax2

  dg = c(dZdx1(old_iteration), dZdx2(old_iteration))
  dgprime = dg*c(sigmax1, sigmax2)
  absqsgsquared = sum(dgprime*dgprime)
  xprime = 1/absqsgsquared*(sum(dgprime*c(old_iteration$X1prime,old_iteration$X2prime))
    - Z(old_iteration))*dgprime
  new_iteration = data.frame(X1 = xprime[1]*sigmax1+mux1,X2 = xprime[2]*sigmax2+mux2 )

  beta = sqrt(sum(xprime^2))
  absq = abs(Z(new_iteration))

  print(paste("Iteration:",it,"beta = ",beta,"|Z| = ",absq))

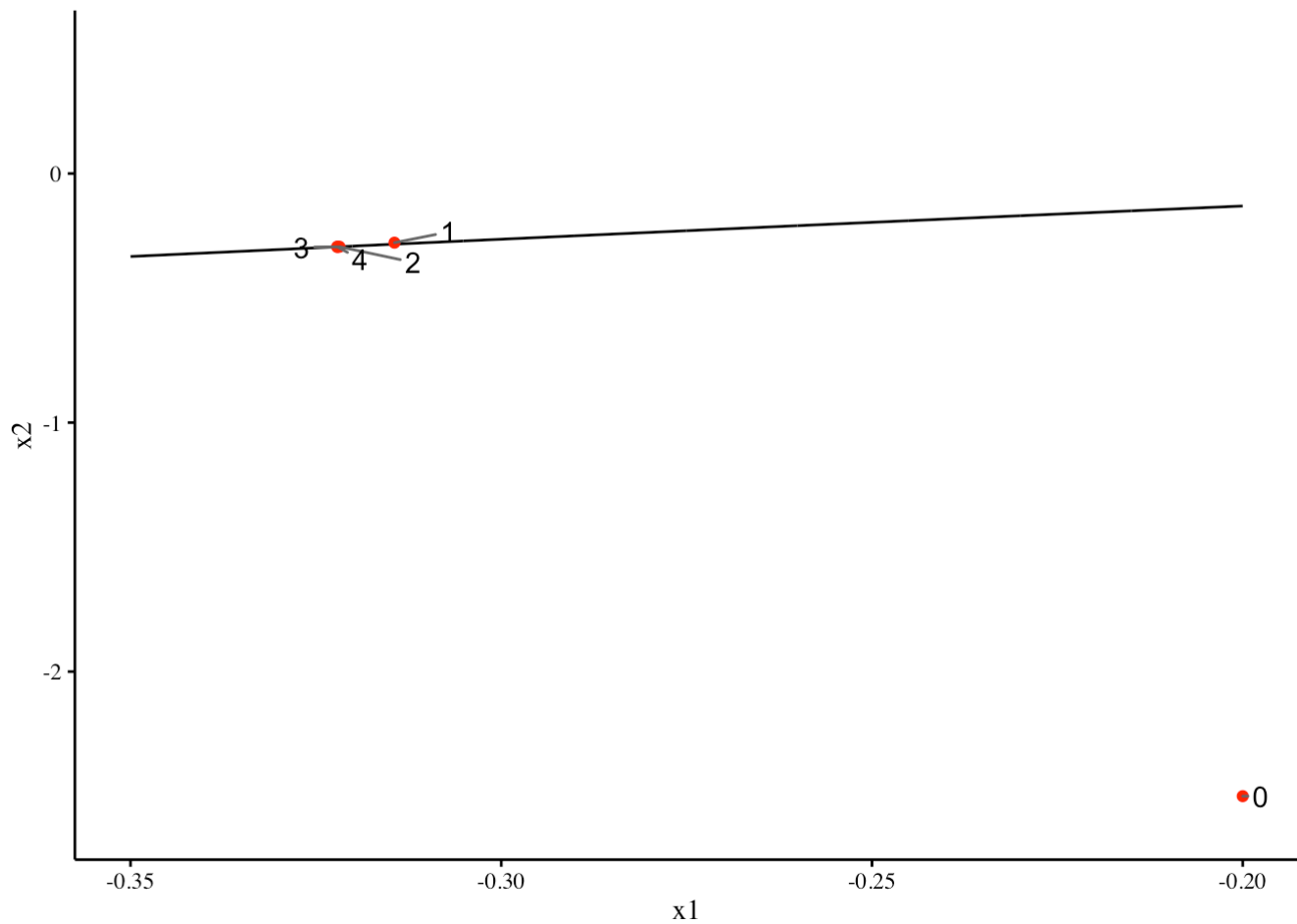
  it = it + 1

  if ( (abs(beta_old-beta) < delta) & (absq < delta))
    converged = 1
  if (it > 10)
    converged = 1
  beta_old = beta
  old_iteration = new_iteration
  new_iteration$lab = it-1
  allits = rbind(allits,new_iteration)
}
```

```
## [1] "Iteration: 2 beta = 0.915482236146609 |Z| = 1.31604513333196e-05"
## [1] "Iteration: 3 beta = 0.915476970253062 |Z| = 1.82026514833344e-08"
## [1] "Iteration: 4 beta = 0.915476962979242 |Z| = 2.43176243031051e-11"
```

The iteration procedure looks like this on a graph.

```
mainplot = mainplot + geom_point(data=allits, color="red") + geom_text_repel(data=allits, aes(la
bel=lab))
print(mainplot)
```



The probability of failure is $1 - \Phi(\beta) = 0.1799706$. Let's check that with Monte Carlo.

```
N = 1e6
df <- data.frame(X1 = rnorm(N,mux1, sigmax1),X2 = rnorm(N,mux2, sigmax2) )
df$Z = Z(df)
pfail = length(df$Z[df$Z<0])/N
```

The actual probability of failure is 0.182784. The Hasofer-Lind approximation is off by 1.5391689%.