

# Hasher-Lind Non-Normal

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

$$Z = -\beta_m(\log p + 1 - f)$$

. Let's say that

$$\begin{aligned} \beta_m &\sim \mathcal{N}(.004, .0004), \\ p &\sim \mathcal{B}(36.37, 21.3602), \\ f &\sim \mathcal{B}(5.06, .322979). \end{aligned}$$

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.

```
alphap = 36.37
betap = 21.3602
alphaf = 5.06
betaf = 0.322979

muB = qnorm(p=0.5, mean = 0.004, sd = 0.0004)
mup = qbeta(p=0.5, shape1 = alphap, shape2 = betap)
muf = qbeta(p=0.5, shape1 = alphaf, shape2 = betaf)
```

Next we determine the sigmas, normalize the  $x_i$  and find a search direction.

```

meanBeta = function(alpha, beta) alpha/(beta + alpha)

iteration1 = data.frame(betam = muB, p = meanBeta(alphap, betap),
                        f = meanBeta(alphaf, betaf))
sigmabeta = 0.0004
sigmap = (iteration1$p - mup)/qnorm(pbeta(iteration1$p, shape1 = alphap, shape2 = betap))
sigmaf = (iteration1$f - muf)/qnorm(pbeta(iteration1$f, shape1 = alphaf, shape2 = betaf))

iteration1$betamprime = (iteration1$betam-muB)/sigmabeta
iteration1$pprime = (iteration1$p-mup)/sigmap
iteration1$fprime = (iteration1$f-muf)/sigmaf

dg = c(dZdbeta(iteration1), dZdp(iteration1), dZdf(iteration1))
dgprime = dg*c(sigmabeta, sigmap, sigmaf)
absgsquared = sum(dgprime*dgprime)
xprime = 1/absgsquared*(sum(dgprime*c(iteration1$betamprime,iteration1$pprime, iteration1$fprime))
                    - Z(iteration1))*dgprime
iteration2 = data.frame(betam = xprime[1]*sigmabeta+muB, p = xprime[2]*sigmap+mup,
                        f = xprime[3]*sigmaf+muf)

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

print(paste("Iteration:",1,"beta = ",beta,"|Z| = ",absg))

```

```
## [1] "Iteration: 1 beta = 3.29127904099938 |Z| = 0.000242939922615366"
```

```

iteration2$lab = "1"
allits = iteration2

```

After one iteration  $\beta = 3.291279$ , and  $|Z| = 2.4293992 \times 10^{-4}$ .

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)){
  sigmabeta = 0.0004
  sigmap = (old_iteration$p - mup)/qnorm(pbeta(old_iteration$p, shape1 = 36.37, shape2 = 21.3602))
  sigmaf = (old_iteration$f - muf)/qnorm(pbeta(old_iteration$f, shape1 = 5.06, shape2 = .322979))

  old_iteration$betamprime = (old_iteration$betam-muB)/sigmabeta
  old_iteration$pprime = (old_iteration$p-mup)/sigmap
  old_iteration$fprime = (old_iteration$f-muf)/sigmaf

  dg = c(dZdbeta(old_iteration), dZdp(old_iteration), dZdf(old_iteration))
  dgprime = dg*c(sigmabeta, sigmap, sigmaf)
  absquared = sum(dgprime*dgprime)
  xprime = 1/absquared*(sum(dgprime*c(old_iteration$betamprime,old_iteration$pprime, old_iterati
on$fprime))
              - Z(old_iteration))*dgprime
  new_iteration = data.frame(betam = xprime[1]*sigmabeta+muB, p = xprime[2]*sigmap+mup,
                             f = xprime[3]*sigmaf+muf)

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

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

  it = it + 1

  if ( (abs(beta_old-beta) < delta) & (absg < 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)
}

```

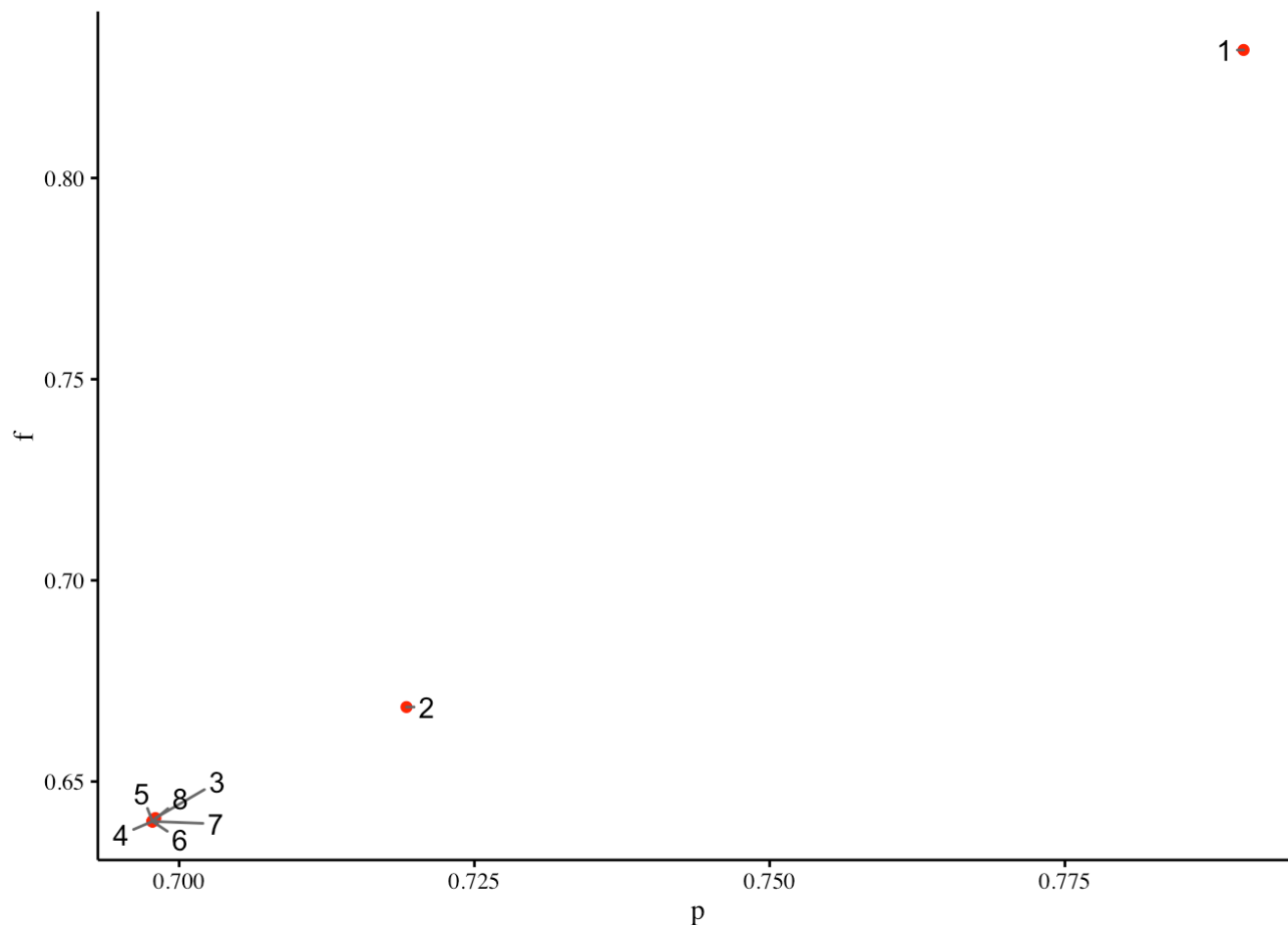
```

## [1] "Iteration: 2 beta = 2.97423430342776 |Z| = 7.62517751970462e-06"
## [1] "Iteration: 3 beta = 2.35802911045196 |Z| = 1.90718926935838e-06"
## [1] "Iteration: 4 beta = 2.29874007184125 |Z| = 7.48846899465322e-10"
## [1] "Iteration: 5 beta = 2.29727313582944 |Z| = 1.25680076830453e-11"
## [1] "Iteration: 6 beta = 2.29710928664281 |Z| = 5.79358782950245e-15"
## [1] "Iteration: 7 beta = 2.29710577663873 |Z| = 2.22044604924992e-18"
## [1] "Iteration: 8 beta = 2.29710570180354 |Z| = 4.44089209850063e-19"

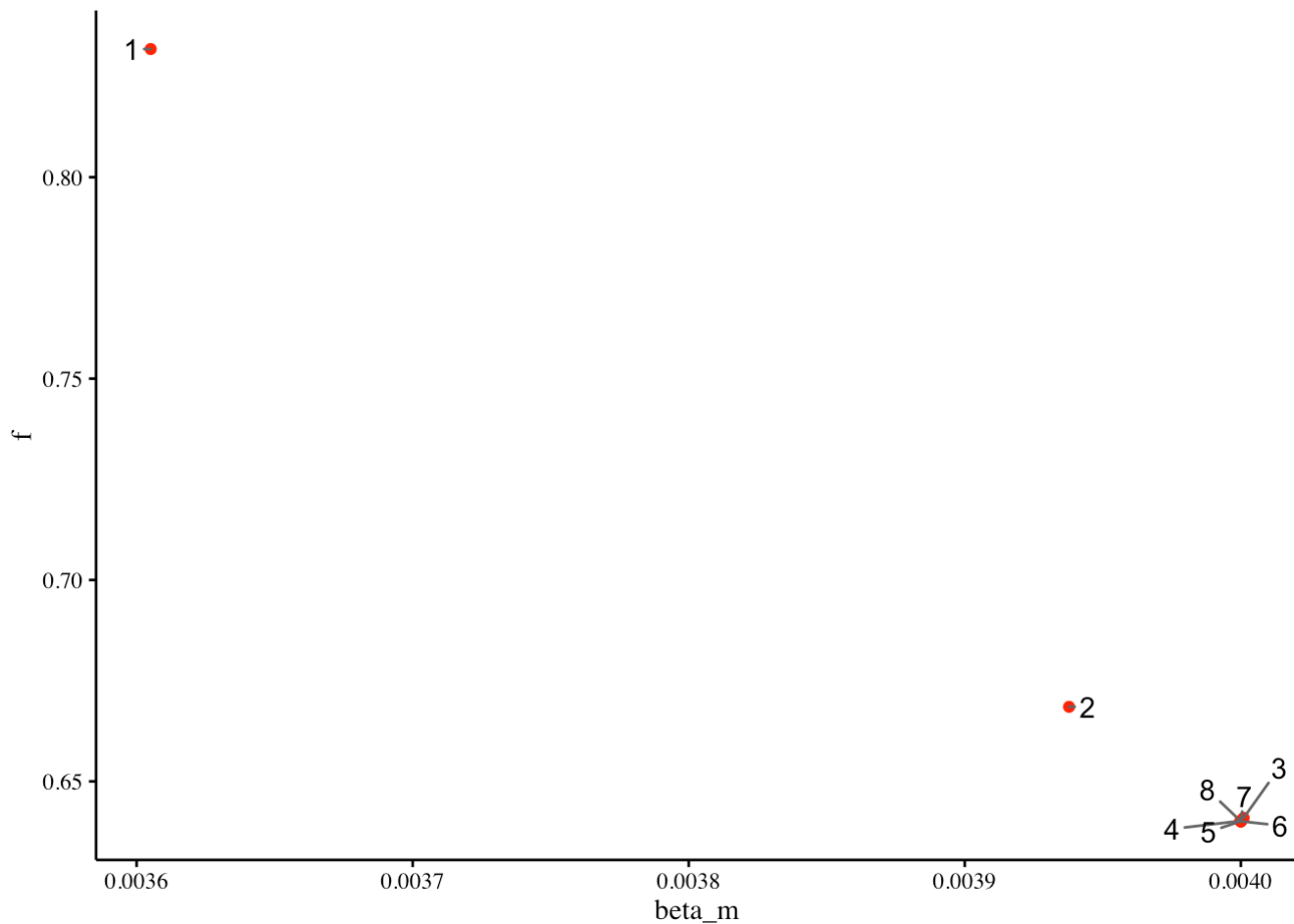
```

The iteration procedure looks like this on a graph.

```
ggplot(allits, aes(x=p, y=f)) + geom_point(color="red") + geom_text_repel(data=allits, aes(label=lab)) + theme_tufte() + theme(axis.line = element_line(color="red"), axis.line.x = element_line(size = .5, colour = "black"), axis.line.y=element_line(size = .5, colour = "black"))
```



```
ggplot(allits, aes(x=betam, y=f)) + geom_point(color="red") + geom_text_repel(data=allits, aes(label=lab)) + scale_x_continuous(name="beta_m") + theme_tufte() + theme(axis.line = element_line(color="red"), axis.line.x = element_line(size = .5, colour = "black"), axis.line.y=element_line(size = .5, colour = "black"))
```



The probability of failure is  $1 - \Phi(\beta) = 0.0108064$ . Let's check that with Monte Carlo.

```
N = 1e6
df <- data.frame(betam = rnorm(N, mean = 0.004, sd = 0.0004),
                 p = rbeta(N, shape1 = alphap, shape2 = betap),
                 f = rbeta(N, shape1 = alphaf, shape2 = betaf) )
df$Z = Z(df)
pfail = length(df$Z[df$Z<0])/N
```

The actual probability of failure is 0.011106. The Hasofer-Lind approximation is off by 2.6979084%.