Distance to Default Analysis

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In this part, I apply Merton's technique for estimating distance-to-default (DD) as well as probability of default (using normal CDF) for a given company.

First I create a Merton function to calculate the distance-to-default.

```
warning=FALSE
library('quantmod')
## Warning: package 'quantmod' was built under R version 4.3.2
## Loading required package: xts
## Warning: package 'xts' was built under R version 4.3.3
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
## Loading required package: TTR
## Warning: package 'TTR' was built under R version 4.3.3
## Registered S3 method overwritten by 'quantmod':
##
     method
                        from
     as.zoo.data.frame zoo
Merton<-function(par,E0,sigmaE,r,T,D,lower,upper)</pre>
  #function to be minimized
  F2plusG2<-function(par,E0,sigmaE,r,T,D)
    A0<-par[1]
    sigmaA<-par[2]</pre>
    d1 < -(log(A0/D) + (r + sigmaA^2/2) *T) / (sigmaA * sqrt(T))
    d2<-d1-sigmaA*sqrt(T)</pre>
    return((E0-A0*pnorm(d1)+exp(-r*T)*D*pnorm(d2))^2
           +(sigmaE*E0-pnorm(d1)*sigmaA*A0)^2)
  # par contains initial values
  result<-optim(par=par, fn=F2plusG2, gr = NULL,</pre>
```

E0=E0, sigmaE=sigmaE, r=r, T=T, D=D,

```
method= "L-BFGS-B", lower=lower, upper=upper)
  return(result)
}
Then, I find the relevant data for ABDE from the Finance dataset.
mystart<-'2023-01-31'
myend <- '2024-02-01' #to get data that ends on the last day of January
myticker<-'ADBE'
stockdata<-getSymbols(Symbols=myticker,</pre>
                       from=mystart,to=myend,auto.assign = FALSE)
head(stockdata)
##
              ADBE.Open ADBE.High ADBE.Low ADBE.Close ADBE.Volume ADBE.Adjusted
                                                                             370.34
## 2023-01-31
                 364.71
                            370.70
                                      364.01
                                                 370.34
                                                             2572400
## 2023-02-01
                 370.01
                            386.72
                                      366.80
                                                 383.92
                                                             3362000
                                                                             383.92
## 2023-02-02
                  393.28
                            402.49
                                      388.88
                                                 392.23
                                                             4020000
                                                                             392.23
                 384.29
                            386.72
## 2023-02-03
                                      377.92
                                                 379.33
                                                             2695400
                                                                             379.33
## 2023-02-06
                  376.21
                            379.29
                                      373.39
                                                 375.23
                                                             2369600
                                                                             375.23
                                                                             383.82
## 2023-02-07
                 373.43
                            384.94
                                      372.76
                                                 383.82
                                                             2476200
Here I find out the annual stock volatility.
sigmaE<-sd(diff(log(stockdata[,6])),na.rm=TRUE)*sqrt(252)</pre>
sigmaE<-round(x=sigmaE,digits=2)</pre>
sigmaE
## [1] 0.32
Relevant data extracted from WRDS.
shortterm < -314 + 73
longterm<-4007
D<-shortterm+0.5*longterm
## [1] 2390.5
shares<-455
PO<-as.numeric(tail(stockdata[,6],1))
E0<-round(P0*shares,2)</pre>
P0
## [1] 617.78
I get the risk free rate from FRED's pre-installed library.
myrate<-getSymbols(Symbols='DGS3MO',src='FRED',auto.assign = FALSE)</pre>
myrate[time(myrate) =='2024-01-31']
              DGS3MO
## 2024-01-31
               5.42
```

```
rfree<-5.42*0.01
rfree
## [1] 0.0542
Now, I estimate the initial value of Assets and initial asset volatility.
AOini<-EO+D; sigmaAini<-0.1
myresult<-Merton(par=c(A0ini,sigmaAini),E0=E0,sigmaE=sigmaE,</pre>
                  r=rfree,T=T,D=D,
                  lower=c(A0ini*0.001,0.001),
                  upper=c(A0ini*10,4))
myresult
## $par
## [1] 2.833543e+05 3.174428e-01
##
## $value
## [1] 1.101143e-20
##
## $counts
## function gradient
##
          20
##
## $convergence
## [1] 0
##
## $message
## [1] "CONVERGENCE: REL_REDUCTION_OF_F <= FACTR*EPSMCH"
A0<-round(x=myresult$par[1],digits=2)
sigmaA<-round(x=myresult$par[2],digits=2)</pre>
ΑO
## [1] 283354.3
sigmaA
## [1] 0.32
Now, I calculate the distance to default and the probability of distance to default.
d1 < -(log(A0/D) + (rfree + sigmaA^2/2) *T)/(sigmaA * sqrt(T))
d2<-round(d1-sigmaA*sqrt(T),2)</pre>
prob<-pnorm(-d2)</pre>
d2
## [1] 14.93
prob
## [1] 1.051331e-50
```

Hence, I conclude that this company is safe with a negligible distance to default.