

Project 4

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```
tinytex::install_tinytex()
```

Project 4.

```
hw4 <- read.csv("/Users/user/Desktop/Yonsei/Junior/3-2/Statistical Models in Finance/stockData2.csv", sep=";", as.is=T)

r_hw4 <- (hw4[,-1, 3:ncol(hw4)]-hw4[-nrow(hw4), 3:ncol(hw4)])/(hw4[-nrow(hw4), 3:ncol(hw4)])

lm1_hw4 <- lm(r_hw4$X.GSPC~r_hw4$AAPL, data=r_hw4)

summary(lm1_hw4)
```

1.

```
##
## Call:
## lm(formula = r_hw4$X.GSPC ~ r_hw4$AAPL, data = r_hw4)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.069380 -0.015170  0.002915  0.011725  0.062321
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.002434   0.003656   0.666   0.508
## r_hw4$AAPL   0.218545   0.049826   4.386 5.03e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02726 on 57 degrees of freedom
## Multiple R-squared:  0.2523, Adjusted R-squared:  0.2392
## F-statistic: 19.24 on 1 and 57 DF,  p-value: 5.033e-05

var(lm1_hw4$residuals)

## [1] 0.0007303755

lm1_hw4$coefficients[1]

## (Intercept)
## 0.002433742

length(colnames(r_hw4))
```

```

alpha_hw4 <- vector()
beta_hw4 <- vector()
varepsilon_hw4 <- vector()

i <- 1

GSPC <- as.vector(r_hw4$X.GSPC)

for (i in 2:30) {

  stock_hw4 <- as.vector(as.matrix(r_hw4)[,i])
  lm1_hw4 <- lm(GSPC~stock_hw4)
  alpha_hw4[i-1] <- lm1_hw4$coefficients[1]
  beta_hw4[i-1] <- lm1_hw4$coefficients[2]
  varepsilon_hw4[i-1] <- var(lm1_hw4$residuals)

}

alpha_hw4

## [1] 2.433742e-03 7.079970e-03 2.844231e-03 3.671962e-03 3.928477e-03
## [6] 1.105572e-04 5.412324e-03 3.231804e-03 1.803766e-03 5.489478e-03
## [11] 3.445247e-03 2.784945e-03 6.678528e-03 4.956295e-03 5.347862e-03
## [16] 2.066278e-04 1.632925e-05 -1.624243e-03 1.584299e-03 -1.181404e-03
## [21] 2.931258e-03 3.891358e-03 5.750581e-03 9.945962e-04 8.680688e-03
## [26] 2.230642e-03 1.989167e-03 6.322300e-03 5.525904e-03

beta_hw4

## [1] 0.21854529 0.29948070 0.30328487 0.18016853 0.06894533 0.22010914
## [7] 0.05138147 0.20912890 0.31592612 0.10811806 0.39011250 0.40793233
## [13] 0.37699174 0.28654722 0.22129542 0.56312853 0.56882943 0.46545424
## [19] 0.34953739 0.42893197 0.63930050 0.27252922 0.19664300 0.25753760
## [25] 0.15613888 0.42504797 0.42272293 0.15484676 0.16124345

varepsilon_hw4

## [1] 0.0007303755 0.0006609972 0.0006602044 0.0008538119 0.0008948190
## [6] 0.0006265819 0.0009404646 0.0008472449 0.0008092451 0.0009427909
## [11] 0.0007602445 0.0007064361 0.0006672216 0.0006957511 0.0007967724
## [16] 0.0005030611 0.0004968727 0.0005001168 0.0005799465 0.0005197844
## [21] 0.0007668186 0.0006461242 0.0008486267 0.0006788692 0.0008784223
## [26] 0.0006944820 0.0006261293 0.0008134866 0.0007296983

```

Note that Alibaba didn't have sufficient data, so that I collected only 29 stocks.

2.

```
covmat_hw4 <- cov(r_hw4[-1])
```

```
covmat_hw4
```

##		AAPL	IBM	GOOGL	META	NFLX
## AAPL		5.161402e-03	8.154776e-04	0.0012866042	1.424916e-03	2.563224e-03
## IBM		8.154776e-04	3.522155e-03	0.0009113984	2.085421e-04	1.636867e-03
## GOOGL		1.286604e-03	9.113984e-04	0.0034429695	2.024376e-03	3.900137e-03
## META		1.424916e-03	2.085421e-04	0.0020243764	3.791748e-03	2.978820e-03
## NFLX		2.563224e-03	1.636867e-03	0.0039001369	2.978820e-03	1.726651e-02
## AMZN		1.902617e-03	1.828950e-03	0.0033756943	2.249842e-03	6.554263e-03
## TSLA		1.122050e-03	4.157924e-04	0.0007026942	1.659269e-03	4.010647e-03
## NKE		5.166174e-04	6.599151e-04	0.0008425891	1.291272e-03	1.195775e-03
## MCD		3.055658e-04	9.711158e-05	0.0007271631	4.180872e-04	-9.065485e-04
## WMT		5.568008e-04	2.525074e-04	0.0001979783	1.042180e-04	-1.050295e-03
## KO		4.636274e-04	3.429239e-04	0.0008405450	2.109764e-04	4.052922e-04
## PEP		6.388790e-04	4.034710e-04	0.0008670723	1.926557e-04	-3.014354e-04
## XOM		1.061794e-04	9.096249e-04	0.0002842845	2.058189e-04	8.020140e-05
## CVX		-1.836160e-04	1.432372e-03	0.0002884520	1.522793e-04	-8.766992e-05
## SHEL		-1.773665e-05	1.102970e-03	0.0003274993	4.537269e-04	1.057686e-03
## BRK.A		5.411592e-04	1.058980e-03	0.0004096452	-9.917718e-05	-1.769499e-04
## BRK.B		5.955250e-04	1.087811e-03	0.0004090822	-6.462784e-05	-1.161862e-04
## V		1.494399e-03	5.255194e-04	0.0018280038	1.475847e-03	2.506864e-03
## JPM		4.542262e-04	1.550686e-03	0.0006922671	6.807921e-05	6.214435e-04
## MA		1.816320e-03	8.865423e-04	0.0015778624	1.218052e-03	2.980241e-03
## C.PJ		7.236398e-04	1.363505e-04	0.0005669395	3.964828e-04	5.079095e-04
## MS		7.004060e-04	1.407620e-03	0.0006506278	2.782306e-04	1.181140e-03
## HSBC		-3.867845e-04	8.620732e-04	0.0005775490	4.156278e-04	2.041948e-03
## BA		1.219026e-03	8.380009e-04	0.0014216121	1.165737e-03	3.327334e-03
## GE		1.266879e-03	4.458322e-04	0.0004676958	6.789714e-04	4.377790e-04
## JNJ		5.154461e-04	6.351447e-04	0.0005315028	7.262320e-05	-1.141233e-03
## PFE		2.332502e-04	9.282776e-04	0.0006974582	6.997456e-05	-6.764219e-04
## PKX		1.229960e-03	1.601849e-03	0.0004760450	1.440172e-03	1.299825e-03
## BIDU		1.970496e-03	7.169175e-04	0.0016521173	2.946421e-03	3.229958e-03
##		AMZN	TSLA	NKE	MCD	WMT
## AAPL		0.0019026173	1.122050e-03	5.166174e-04	3.055658e-04	5.568008e-04
## IBM		0.0018289502	4.157924e-04	6.599151e-04	9.711158e-05	2.525074e-04
## GOOGL		0.0033756943	7.026942e-04	8.425891e-04	7.271631e-04	1.979783e-04
## META		0.0022498417	1.659269e-03	1.291272e-03	4.180872e-04	1.042180e-04
## NFLX		0.0065542633	4.010647e-03	1.195775e-03	-9.065485e-04	-1.050295e-03
## AMZN		0.0072306900	1.557785e-03	1.382206e-03	8.401319e-04	3.276239e-04
## TSLA		0.0015577852	1.379896e-02	2.569594e-04	3.655092e-04	4.146595e-04
## NKE		0.0013822056	2.569594e-04	2.964446e-03	2.633847e-04	2.890924e-04
## MCD		0.0008401319	3.655092e-04	2.633847e-04	1.679698e-03	7.307442e-04
## WMT		0.0003276239	4.146595e-04	2.890924e-04	7.307442e-04	2.917464e-03
## KO		0.0008277755	9.833069e-04	2.500265e-04	9.297492e-04	8.259870e-04
## PEP		0.0009116359	5.669634e-04	4.484398e-04	7.122870e-04	6.440811e-04
## XOM		0.0005998186	5.546061e-05	3.469047e-04	6.200364e-04	2.226793e-04
## CVX		0.0004753342	1.703241e-04	-2.245262e-06	9.045952e-04	-1.107870e-04
## SHEL		0.0008604151	5.767812e-04	3.589788e-05	6.554628e-04	1.094697e-04
## BRK.A		0.0007943267	5.205452e-05	4.805429e-04	5.338279e-04	6.546147e-04
## BRK.B		0.0008115421	1.243223e-04	4.975719e-04	5.470123e-04	6.293613e-04
## V		0.0026514995	6.745652e-04	1.000524e-03	5.577183e-04	2.315395e-04

##	JPM	0.0011990638	7.633182e-04	6.925872e-04	4.975185e-04	-4.101389e-06
##	MA	0.0025345065	9.154285e-04	7.795380e-04	3.145465e-04	-1.059468e-04
##	C.PJ	0.0007828357	4.861594e-04	9.052776e-05	1.640488e-04	8.709971e-05
##	MS	0.0010076453	1.031604e-03	5.275087e-04	1.992203e-04	-1.915928e-06
##	HSBC	0.0016235425	1.384753e-03	1.541834e-04	3.331698e-04	-5.400316e-05
##	BA	0.0025565115	3.122172e-04	9.459638e-04	6.299939e-04	3.978525e-04
##	GE	0.0004048987	6.462503e-04	2.669322e-04	4.074932e-04	-3.853461e-04
##	JNJ	0.0006767991	3.924255e-04	3.050479e-04	8.862411e-04	7.709756e-04
##	PFE	0.0015203212	2.729686e-04	5.098308e-04	5.996435e-04	3.522962e-04
##	PKX	0.0013885741	7.089264e-04	-3.053373e-05	6.380441e-04	4.073173e-04
##	BIDU	0.0027070303	8.594939e-04	1.103980e-03	6.597535e-04	-7.497932e-04
##		KO	PEP	XOM	CVX	SHEL
##	AAPL	4.636274e-04	6.388790e-04	1.061794e-04	-1.836160e-04	-1.773665e-05
##	IBM	3.429239e-04	4.034710e-04	9.096249e-04	1.432372e-03	1.102970e-03
##	GOOGL	8.405450e-04	8.670723e-04	2.842845e-04	2.884520e-04	3.274993e-04
##	META	2.109764e-04	1.926557e-04	2.058189e-04	1.522793e-04	4.537269e-04
##	NFLX	4.052922e-04	-3.014354e-04	8.020140e-05	-8.766992e-05	1.057686e-03
##	AMZN	8.277755e-04	9.116359e-04	5.998186e-04	4.753342e-04	8.604151e-04
##	TSLA	9.833069e-04	5.669634e-04	5.546061e-05	1.703241e-04	5.767812e-04
##	NKE	2.500265e-04	4.484398e-04	3.469047e-04	-2.245262e-06	3.589788e-05
##	MCD	9.297492e-04	7.122870e-04	6.200364e-04	9.045952e-04	6.554628e-04
##	WMT	8.259870e-04	6.440811e-04	2.226793e-04	-1.107870e-04	1.094697e-04
##	KO	1.423571e-03	1.050509e-03	5.534537e-04	6.427843e-04	7.412262e-04
##	PEP	1.050509e-03	1.625266e-03	6.041965e-04	7.230605e-04	5.075905e-04
##	XOM	5.534537e-04	6.041965e-04	2.178912e-03	2.150856e-03	2.052789e-03
##	CVX	6.427843e-04	7.230605e-04	2.150856e-03	3.424015e-03	2.611105e-03
##	SHEL	7.412262e-04	5.075905e-04	2.052789e-03	2.611105e-03	3.678092e-03
##	BRK.A	6.417387e-04	6.243327e-04	6.701885e-04	7.635481e-04	6.919924e-04
##	BRK.B	6.455710e-04	6.248171e-04	6.675790e-04	7.580065e-04	6.905998e-04
##	V	5.440545e-04	6.595379e-04	3.713907e-04	3.830234e-04	3.244132e-04
##	JPM	2.775633e-04	3.678218e-04	8.777800e-04	1.383463e-03	6.996301e-04
##	MA	3.545187e-04	5.455099e-04	5.387545e-04	6.859243e-04	5.850646e-04
##	C.PJ	2.342371e-04	3.492504e-04	4.494090e-05	1.309495e-04	5.142941e-05
##	MS	5.010228e-05	9.336812e-05	7.874916e-04	1.399439e-03	8.427665e-04
##	HSBC	3.579933e-04	3.398898e-04	8.003459e-04	1.441499e-03	1.369684e-03
##	BA	7.786148e-04	6.730496e-04	7.377771e-04	1.134168e-03	1.393781e-03
##	GE	4.790724e-04	3.087148e-04	8.837439e-04	1.211094e-03	1.092346e-03
##	JNJ	9.419922e-04	1.131900e-03	8.046473e-04	8.559758e-04	8.056109e-04
##	PFE	4.969867e-04	7.325690e-04	7.386476e-04	9.088495e-04	6.677733e-04
##	PKX	5.396295e-04	6.388812e-04	1.024243e-03	2.142115e-03	2.590820e-03
##	BIDU	6.171448e-04	6.553895e-04	1.605039e-03	2.499057e-03	2.037649e-03
##		BRK.A	BRK.B	V	JPM	MA
##	AAPL	5.411592e-04	5.955250e-04	0.0014943987	4.542262e-04	0.0018163197
##	IBM	1.058980e-03	1.087811e-03	0.0005255194	1.550686e-03	0.0008865423
##	GOOGL	4.096452e-04	4.090822e-04	0.0018280038	6.922671e-04	0.0015778624
##	META	-9.917718e-05	-6.462784e-05	0.0014758471	6.807921e-05	0.0012180523
##	NFLX	-1.769499e-04	-1.161862e-04	0.0025068639	6.214435e-04	0.0029802406
##	AMZN	7.943267e-04	8.115421e-04	0.0026514995	1.199064e-03	0.0025345065
##	TSLA	5.205452e-05	1.243223e-04	0.0006745652	7.633182e-04	0.0009154285
##	NKE	4.805429e-04	4.975719e-04	0.0010005244	6.925872e-04	0.0007795380
##	MCD	5.338279e-04	5.470123e-04	0.0005577183	4.975185e-04	0.0003145465
##	WMT	6.546147e-04	6.293613e-04	0.0002315395	-4.101389e-06	-0.0001059468
##	KO	6.417387e-04	6.455710e-04	0.0005440545	2.775633e-04	0.0003545187
##	PEP	6.243327e-04	6.248171e-04	0.0006595379	3.678218e-04	0.0005455099

##	XOM	6.701885e-04	6.675790e-04	0.0003713907	8.777800e-04	0.0005387545
##	CVX	7.635481e-04	7.580065e-04	0.0003830234	1.383463e-03	0.0006859243
##	SHEL	6.919924e-04	6.905998e-04	0.0003244132	6.996301e-04	0.0005850646
##	BRK.A	1.494207e-03	1.478370e-03	0.0006244556	1.302157e-03	0.0005744461
##	BRK.B	1.478370e-03	1.483532e-03	0.0006564324	1.267561e-03	0.0006025958
##	V	6.244556e-04	6.564324e-04	0.0022007061	8.729463e-04	0.0019838292
##	JPM	1.302157e-03	1.267561e-03	0.0008729463	3.248976e-03	0.0010395920
##	MA	5.744461e-04	6.025958e-04	0.0019838292	1.039592e-03	0.0024845295
##	C.PJ	1.765480e-04	1.845724e-04	0.0004715384	1.886344e-04	0.0004423973
##	MS	1.448726e-03	1.382435e-03	0.0008500588	3.249996e-03	0.0011034799
##	HSBC	5.277662e-04	5.162097e-04	0.0007268910	1.198023e-03	0.0008571328
##	BA	1.176696e-03	1.206026e-03	0.0012919606	1.250543e-03	0.0015172625
##	GE	3.345867e-04	2.970479e-04	0.0002909159	7.560151e-04	0.0003469740
##	JNJ	7.291119e-04	7.519546e-04	0.0004663965	7.275413e-04	0.0002432757
##	PFE	8.113435e-04	8.375196e-04	0.0008852594	1.237453e-03	0.0006969986
##	PKX	1.221256e-03	1.132196e-03	0.0007612663	8.685071e-04	0.0012328237
##	BIDU	7.371205e-04	7.243659e-04	0.0022213311	1.131771e-03	0.0022152715
##	C.PJ		MS	HSBC	BA	GE
##	AAPL	7.236398e-04	7.004060e-04	-3.867845e-04	0.0012190263	0.0012668788
##	IBM	1.363505e-04	1.407620e-03	8.620732e-04	0.0008380009	0.0004458322
##	GOOGL	5.669395e-04	6.506278e-04	5.775490e-04	0.0014216121	0.0004676958
##	META	3.964828e-04	2.782306e-04	4.156278e-04	0.0011657374	0.0006789714
##	NFLX	5.079095e-04	1.181140e-03	2.041948e-03	0.0033273344	0.0004377790
##	AMZN	7.828357e-04	1.007645e-03	1.623542e-03	0.0025565115	0.0004048987
##	TSLA	4.861594e-04	1.031604e-03	1.384753e-03	0.0003122172	0.0006462503
##	NKE	9.052776e-05	5.275087e-04	1.541834e-04	0.0009459638	0.0002669322
##	MCD	1.640488e-04	1.992203e-04	3.331698e-04	0.0006299939	0.0004074932
##	WMT	8.709971e-05	-1.915928e-06	-5.400316e-05	0.0003978525	-0.0003853461
##	KO	2.342371e-04	5.010228e-05	3.579933e-04	0.0007786148	0.0004790724
##	PEP	3.492504e-04	9.336812e-05	3.398898e-04	0.0006730496	0.0003087148
##	XOM	4.494090e-05	7.874916e-04	8.003459e-04	0.0007377771	0.0008837439
##	CVX	1.309495e-04	1.399439e-03	1.441499e-03	0.0011341677	0.0012110945
##	SHEL	5.142941e-05	8.427665e-04	1.369684e-03	0.0013937813	0.0010923460
##	BRK.A	1.765480e-04	1.448726e-03	5.277662e-04	0.0011766960	0.0003345867
##	BRK.B	1.845724e-04	1.382435e-03	5.162097e-04	0.0012060259	0.0002970479
##	V	4.715384e-04	8.500588e-04	7.268910e-04	0.0012919606	0.0002909159
##	JPM	1.886344e-04	3.249996e-03	1.198023e-03	0.0012505432	0.0007560151
##	MA	4.423973e-04	1.103480e-03	8.571328e-04	0.0015172625	0.0003469740
##	C.PJ	5.140040e-04	3.256405e-04	2.979412e-04	0.0003567673	0.0004188118
##	MS	3.256405e-04	4.453491e-03	1.880611e-03	0.0017847081	0.0008687923
##	HSBC	2.979412e-04	1.880611e-03	3.317119e-03	0.0013099674	0.0005850154
##	BA	3.567673e-04	1.784708e-03	1.309967e-03	0.0044933693	0.0004780132
##	GE	4.188118e-04	8.687923e-04	5.850154e-04	0.0004780132	0.0040391680
##	JNJ	2.974854e-04	3.850925e-04	1.375320e-04	0.0005181394	0.0002600534
##	PFE	3.258579e-04	1.028323e-03	3.958526e-04	0.0010737447	0.0002251642
##	PKX	3.808461e-04	1.330097e-03	1.768446e-03	0.0018717670	0.0012291192
##	BIDU	4.081753e-04	1.990092e-03	2.153650e-03	0.0030065042	0.0013022453
##	JNJ		PFE	PKX	BIDU	
##	AAPL	0.0005154461	2.332502e-04	1.229960e-03	0.0019704964	
##	IBM	0.0006351447	9.282776e-04	1.601849e-03	0.0007169175	
##	GOOGL	0.0005315028	6.974582e-04	4.760450e-04	0.0016521173	
##	META	0.0000726232	6.997456e-05	1.440172e-03	0.0029464214	
##	NFLX	-0.0011412327	-6.764219e-04	1.299825e-03	0.0032299583	
##	AMZN	0.0006767991	1.520321e-03	1.388574e-03	0.0027070303	

## TSLA	0.0003924255	2.729686e-04	7.089264e-04	0.0008594939
## NKE	0.0003050479	5.098308e-04	-3.053373e-05	0.0011039798
## MCD	0.0008862411	5.996435e-04	6.380441e-04	0.0006597535
## WMT	0.0007709756	3.522962e-04	4.073173e-04	-0.0007497932
## KO	0.0009419922	4.969867e-04	5.396295e-04	0.0006171448
## PEP	0.0011318998	7.325690e-04	6.388812e-04	0.0006553895
## XOM	0.0008046473	7.386476e-04	1.024243e-03	0.0016050390
## CVX	0.0008559758	9.088495e-04	2.142115e-03	0.0024990572
## SHEL	0.0008056109	6.677733e-04	2.590820e-03	0.0020376493
## BRK.A	0.0007291119	8.113435e-04	1.221256e-03	0.0007371205
## BRK.B	0.0007519546	8.375196e-04	1.132196e-03	0.0007243659
## V	0.0004663965	8.852594e-04	7.612663e-04	0.0022213311
## JPM	0.0007275413	1.237453e-03	8.685071e-04	0.0011317707
## MA	0.0002432757	6.969986e-04	1.232824e-03	0.0022152715
## C.PJ	0.0002974854	3.258579e-04	3.808461e-04	0.0004081753
## MS	0.0003850925	1.028323e-03	1.330097e-03	0.0019900916
## HSBC	0.0001375320	3.958526e-04	1.768446e-03	0.0021536500
## BA	0.0005181394	1.073745e-03	1.871767e-03	0.0030065042
## GE	0.0002600534	2.251642e-04	1.229119e-03	0.0013022453
## JNJ	0.0015631773	9.741587e-04	7.078725e-04	0.0005635115
## PFE	0.0009741587	1.962931e-03	7.021218e-06	0.0004981744
## PKX	0.0007078725	7.021218e-06	6.815053e-03	0.0033531749
## BIDU	0.0005635115	4.981744e-04	3.353175e-03	0.0095077521

3.

```
means_hw4 <- colMeans(r_hw4[-1])
corrmat_hw4 <- cor(r_hw4[-1])
variances_hw4 <- diag(covmat_hw4)
stdev_hw4 <- diag(covmat_hw4)^(0.5)

B2_hw4 <- t(means_hw4) %*% covmat_hw4^(-1) %*% means_hw4

A_hw4 <- sum(covmat_hw4^(-1) * means_hw4)
B_hw4 <- sum(covmat_hw4^(-1) * means_hw4 * means_hw4)
C_hw4 <- sum(covmat_hw4^(-1))
D_hw4 <- B_hw4 * C_hw4 - A_hw4 * A_hw4

x_hw4 <- seq(-0.5, 0.5, 0.001)
sigma_squared_hw4 <- (C_hw4 * x_hw4 * x_hw4 - 2 * A_hw4 * x_hw4 + B_hw4) / D_hw4

hw2 <- read.csv("/Users/user/Desktop/Yonsei/Junior/3-2/Statistical Models in Finance/stockData.csv", sep=";", as.is=T)

r_hw2 <- (hw2[-1, 3:ncol(hw2)] - hw2[-nrow(hw2), 3:ncol(hw2)]) / hw2[-nrow(hw2), 3:ncol(hw2)]
means_hw2 <- colMeans(r_hw2[-1])
covmat_hw2 <- cov(r_hw2[-1])
corrmat_hw2 <- cor(r_hw2[-1])
variances_hw2 <- diag(covmat_hw2)
stdev_hw2 <- diag(covmat_hw2)^(0.5)

A_hw2 <- sum(covmat_hw2^(-1) * means_hw2)
B_hw2 <- sum(covmat_hw2^(-1) * means_hw2 * means_hw2)
C_hw2 <- sum(covmat_hw2^(-1))
D_hw2 <- B_hw2 * C_hw2 - A_hw2 * A_hw2

B2_hw2 <- t(means_hw2) %*% covmat_hw2^(-1) %*% means_hw2

sigma_squared_hw2 <- (C_hw2 * x_hw4 * x_hw4 - 2 * A_hw2 * x_hw4 + B_hw2) / D_hw2

sigma_squared_hw4_2 <- (C_hw4 * x_hw4 * x_hw4 - 2 * A_hw4 * x_hw4 + B2_hw4) / D_hw4

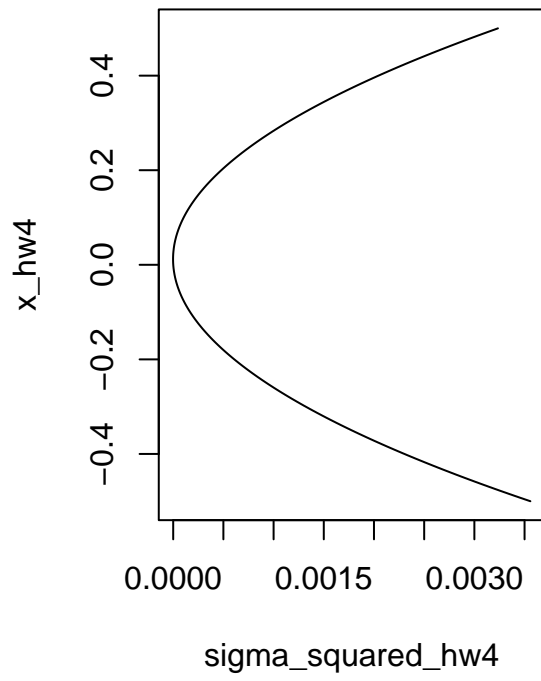
## Warning in C_hw4 * x_hw4 * x_hw4 - 2 * A_hw4 * x_hw4 + B2_hw4: Recycling array of length 1 in vector
## Use c() or as.vector() instead.

sigma_squared_hw2_2 <- (C_hw2 * x_hw4 * x_hw4 - 2 * A_hw2 * x_hw4 + B2_hw2) / D_hw2

## Warning in C_hw2 * x_hw4 * x_hw4 - 2 * A_hw2 * x_hw4 + B2_hw2: Recycling array of length 1 in vector
## Use c() or as.vector() instead.

par(mfrow=c(1,2))
plot(sigma_squared_hw4, x_hw4, type='l', main='2014/01/01-19/01/01')
plot(sigma_squared_hw2, x_hw4, type='l', main='2015/01/01-20/01/01')
```

2014/01/01–19/01/01



2015/01/01–20/01/01

