

Report - Aug 1st

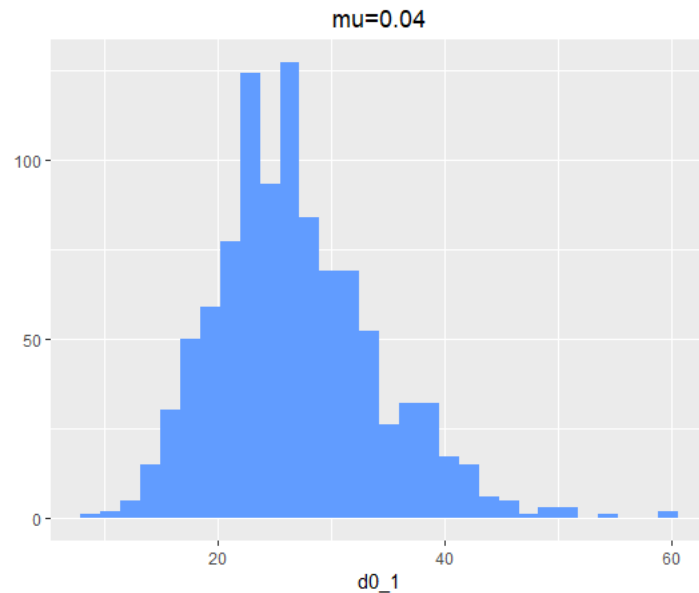
Tiancheng Hou

Note: Initial asset, b_0 , is set as 500

1. Method I: No Recalibration

$\mu=0.04$

distribution of all 1000 d_0



```
> summary(d0_1)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
  9.178  22.062  26.036  26.867  31.025  60.102
```

0.05 quantile:

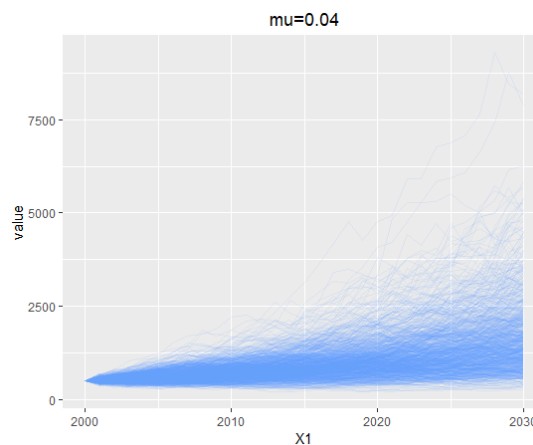
```
> mylist[[1]]$withdrawal
[1] 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628
[13] 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628
[25] 16.55628 16.55628 16.55628 16.55628 16.55628 16.55628
```

Ruin probability:

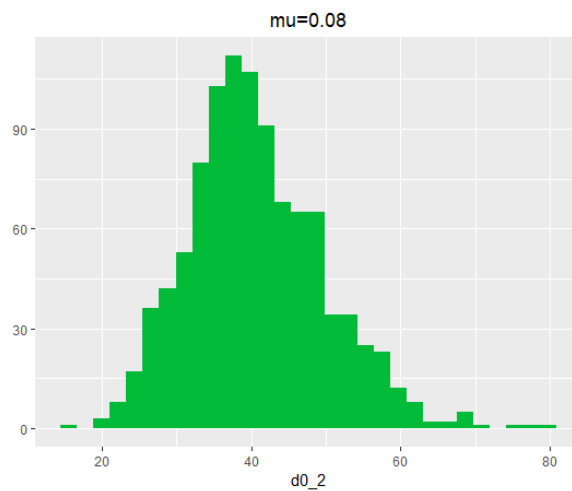
```
> sum(a)/1000
[1] 0.049
```

It should be exactly 0.05, and I get 0.049. I think it is because, for the 50th smallest d_0 , it generates a very small ending value, though very close to 0, slightly greater than 0. (?)

Path of price (Initial price is 500):



mu=0.08



```
> summary(d0_2)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
16.40	34.62	39.71	40.66	46.41	80.61

0.05 quantile:

```
mylist[[1]]$withdrwal
```

[1] 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839

[13] 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839

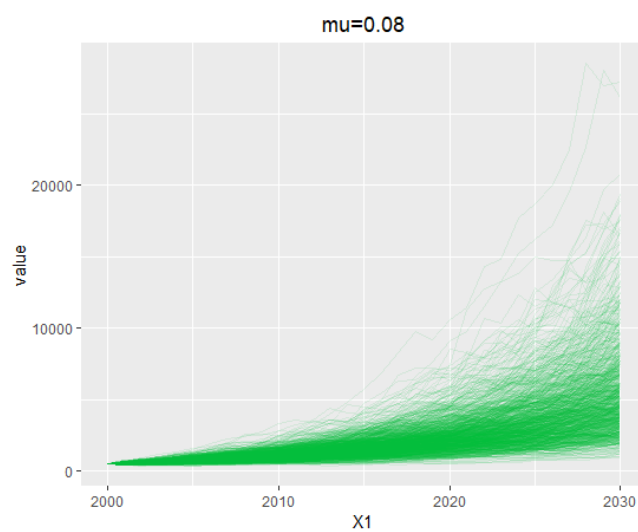
[25] 27.09839 27.09839 27.09839 27.09839 27.09839 27.09839

Ruin probability:

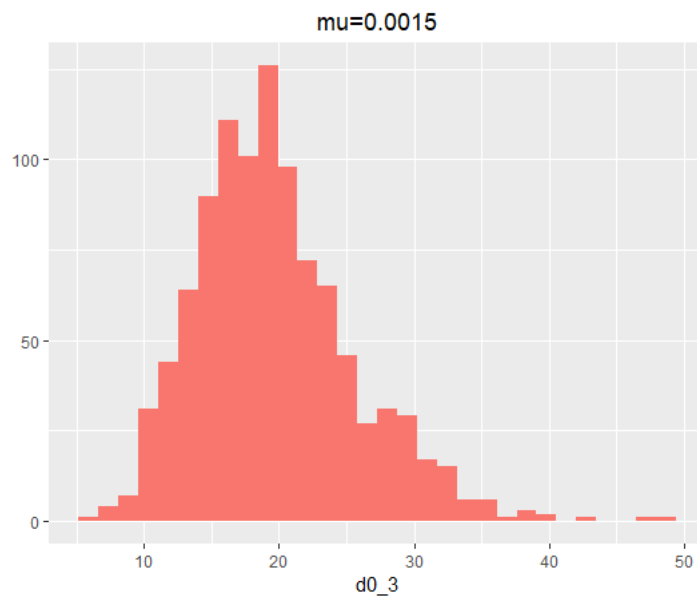
```
> sum(a)/1000
```

[1] 0.573

Path of price (Initial price is 500)



mu=0.015



```
> summary(d0_3)
   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
 6.056 15.649 18.941 19.654 22.879 48.836
```

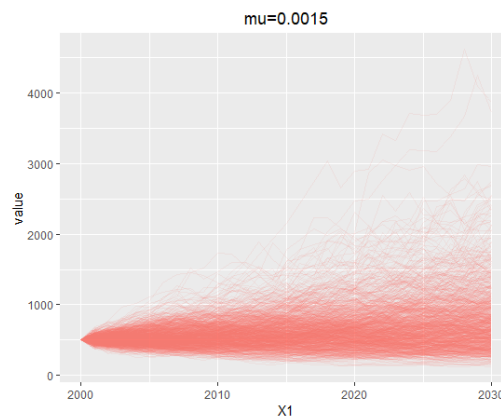
0.05 quantile:

```
> mylist[[1]]$withdwal
 [1] 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182
[13] 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182
[25] 11.50182 11.50182 11.50182 11.50182 11.50182 11.50182
```

Ruin probability:

```
> sum(a)/1000
[1] 0.003
```

Path of price (Initial price is 500)



2. Method II: Kalman Filter Recalibration

b.

Note: Initial μ and σ are randomly set as 0.04 and 0.2, since we do not know the true value of them.

Result:

```
> optimal_kf
$`withdrawal`
[1] 23.28553 27.56753 22.93079 37.09878 54.77432 61.91534 48.24402 34.89147 37.18334 35.37883 34.24792 38.77763
[13] 37.47057 38.91209 41.66017 46.15855 44.42080 50.04995 36.89462 34.17286 33.44282 34.88557 36.81196 35.37293
[25] 37.95235 36.19516 43.31822 45.79551 41.03983 43.74838

$Remaining_value
[1] 471.67800 520.50877 575.14433 642.82535 709.35047 565.12214 410.16770 439.66979 417.74525 393.79844 437.61164
[12] 412.43993 412.21058 422.94969 448.97992 409.70164 445.52915 316.63001 278.83721 255.90341 248.10453 240.73728
[25] 208.68863 197.75838 161.42315 158.93471 130.09573 80.03129 43.74838 0.00000
```

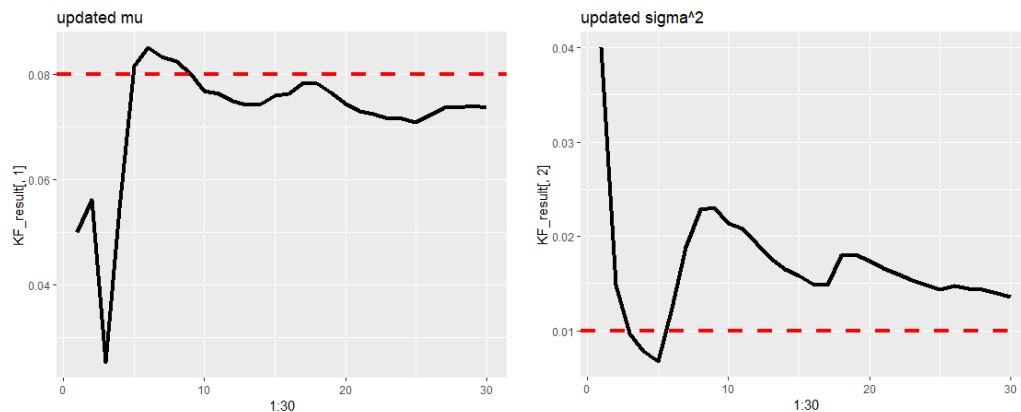
The first vector is the withdrawal amount in each year. The second vector is the remaining value of the founding in each year. (Note: Initial asset, b_0 , is set as 500)

Mean of withdrawal vector is

```
> mean(optimal_kf$`withdrawal`)
[1] 39.15326
```

which is greater than $d_0(23.28553)$. It means that K-F recalibration can reflect the market performance.

Update μ and σ^2 with Kalman filter:



From above, we notice that even we set μ and σ that is not very close to the “actual value”, they will converge to the it, as Kalman filter gathering more data.

c.

Result:

```
> optimal_kf
$withdrawal
[1] 26.33398 24.38049 14.57325 20.01644 27.67902 30.08672 22.54653 15.72576 16.21313 14.92780 14.10588 15.66889
[13] 14.93397 15.28293 16.23366 17.88514 17.11876 19.20607 14.09040 12.97561 12.66250 13.17617 13.88065 13.32303
[25] 14.28137 13.61115 16.28568 17.20832 15.41554 16.43117

$Remaining_value
[1] 439.16768 455.54704 477.63850 512.33233 547.83262 423.48067 298.13791 310.04863 285.78233 261.39737 282.03888
[13] 258.12180 250.51633 249.74817 257.66581 228.52842 241.62671 166.86851 142.70372 127.14823 119.64756 112.64744
[23] 94.71297 87.02390 68.85443 65.70035 52.10387 31.04360 16.43117 0.00000
```

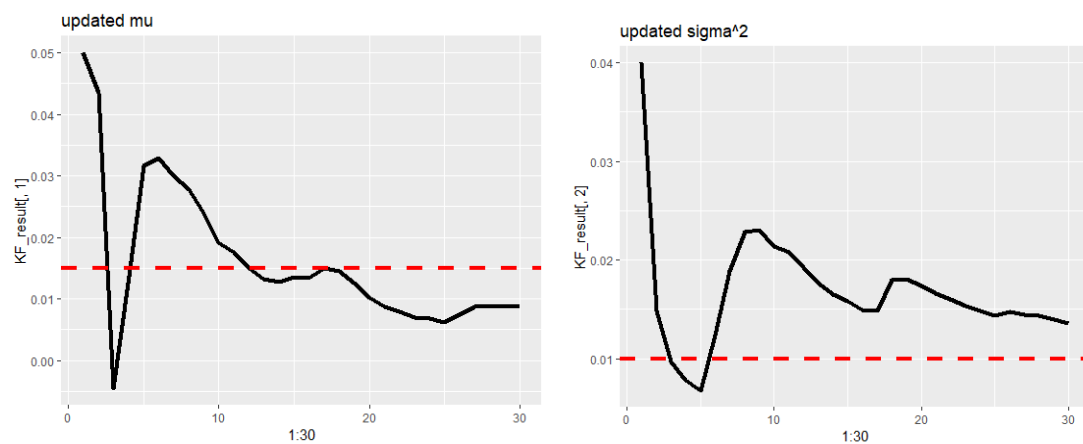
The first vector is the withdrawal amount in each year. The second vector is the remaining value of the founding in each year. (Note: Initial asset, b_0 , is set as 500)

Mean of withdrawal vector is

```
> mean(optimal_kf$withdrawal)
[1] 17.20867
```

which is smaller than $d_0(26.33398)$. It means that K-F recalibration can reflect the market performance.

Update μ and σ^2 with Kalman filter:



From above, we notice that even we set μ and σ^2 that is not very close to the “actual value”, they will converge to the it, as Kalman filter gathering more data.

3. Naïve Recalibration

b.

Result:

```
> optimal_naive
$`withdrwal`
 [1] 16.55628 15.77460 18.89424 21.85478 25.84019 31.18718 26.61348 21.11574 24.14392 24.99108 26.08557
[12] 31.59387 32.57517 35.72315 39.99397 48.62324 50.13624 62.23529 49.19007 49.31723 52.81609 58.83476
[23] 66.37643 68.40674 80.63644 85.13711 112.35189 131.36205 133.66223 170.14691

$Remaining_value
 [1] 478.3362 542.1339 604.8063 696.4764 808.9708 678.8985 517.6196 581.6989 578.6926 570.2561 662.3141 652.2368
[13] 681.2052 731.3354 814.0755 778.5263 888.4096 661.4547 610.2988 586.7512 595.4839 605.9098 552.1359 551.9494
[25] 476.0823 496.1731 431.8890 285.3084 170.1469 0.0000
```

The first vector is the withdrawal amount in each year. The second vector is the remaining value of the founding in each year. (Note: Initial asset, b_0 , is set as 500)

Mean of withdrawal vector is

```
> mean(optimal_naive$withdrwal)
[1] 53.7392
```

which greater than $d_0(16.55628)$.

c.

```
> optimal_naive
$`withdrwal`
 [1] 16.55628 14.78187 16.59095 17.98287 19.92412 22.53359 18.01884 13.39684 14.35405 13.92268 13.61786 15.45547
[13] 14.93265 15.34514 16.09854 18.34032 17.72089 20.61301 15.26697 14.34317 14.39406 15.02526 15.88447 15.34012
[25] 16.94462 16.76449 20.73112 22.71345 21.65672 25.83325

$Remaining_value
 [1] 448.23325 476.04530 497.65562 537.01925 584.50355 459.65292 328.40272 345.83193 322.39305 297.69981 323.99883
[12] 298.98926 292.61669 294.38023 307.06304 275.17376 294.25093 205.29368 177.49610 159.90835 152.07509 144.99963
[23] 123.81570 115.98447 93.74616 91.55362 74.67674 46.22731 25.83325 0.00000
```

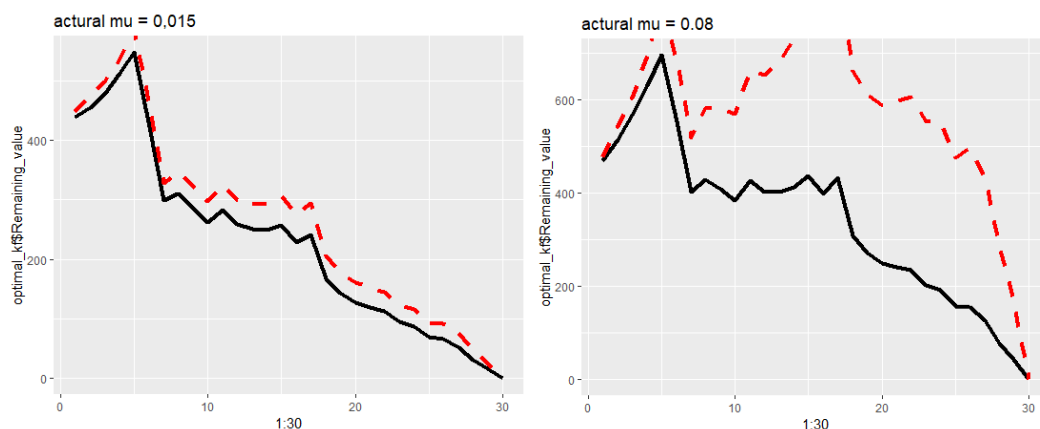
The first vector is the withdrawal amount in each year. The second vector is the remaining value of the founding in each year. (Note: Initial asset, b_0 , is set as 500)

Mean of withdrawal vector is

```
> mean(optimal_naive$withdrwal)
[1] 17.16946
```

which is greater than $d_0(16.55628)$.

Advantage of Kalman filter:



Black curve is the remaining value calculated with updated μ and σ (Kalman), red dash line are remaining value calculated with fixed μ and σ (Naïve). Notice that the black line is much smoother than the red one. It suggests the advantage of Kalman filter.