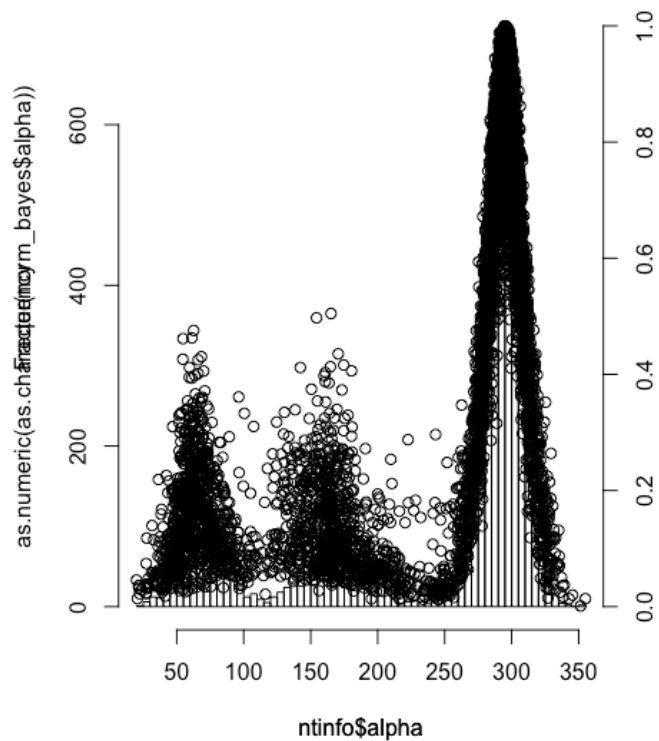
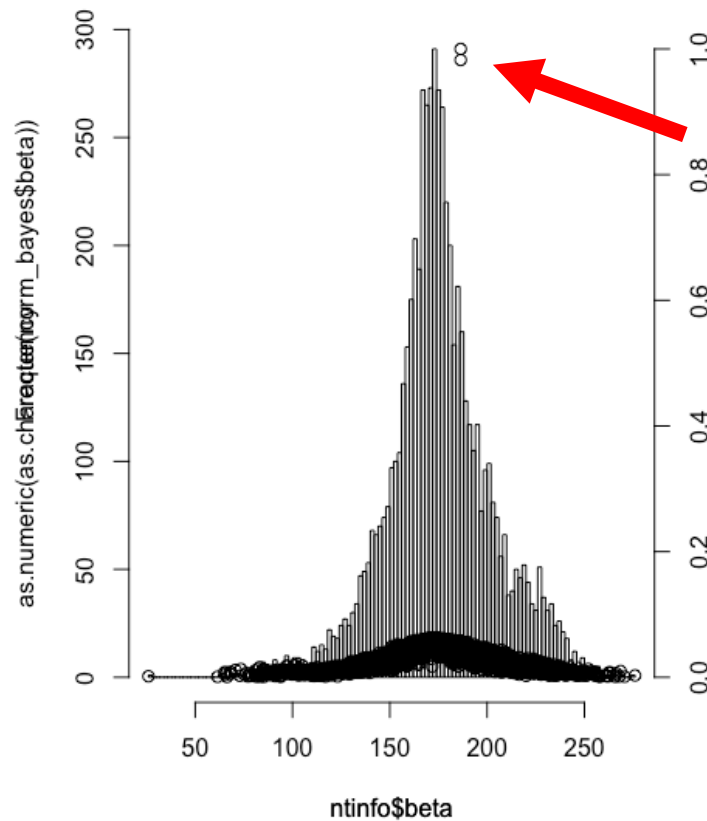


Check all the plots generated

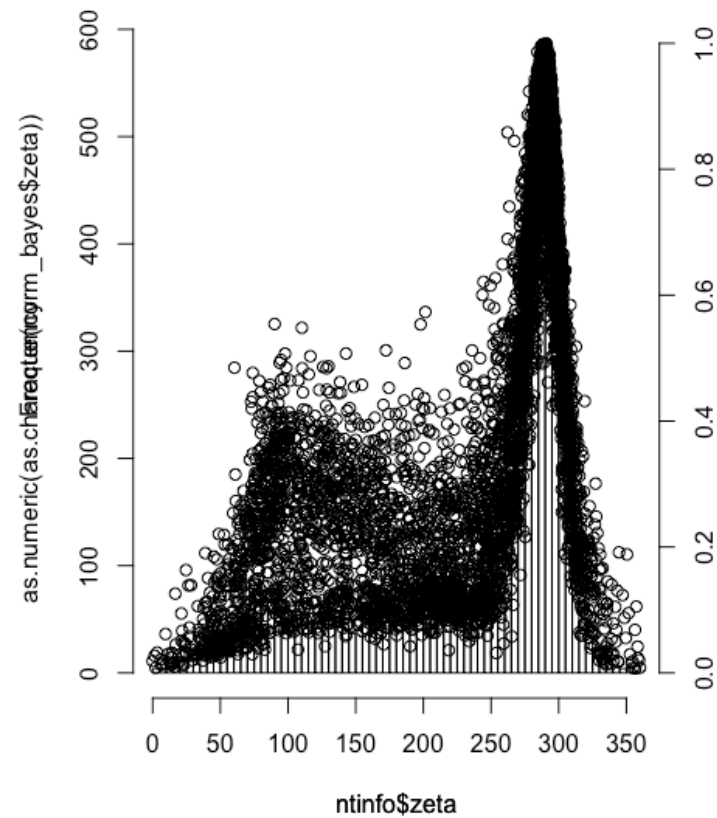
Histogram of ntinfo\$alpha



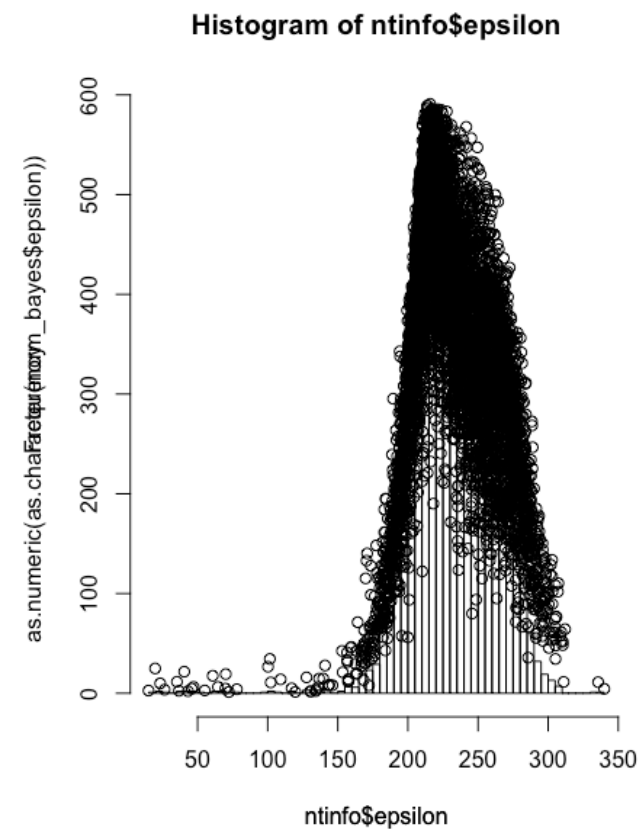
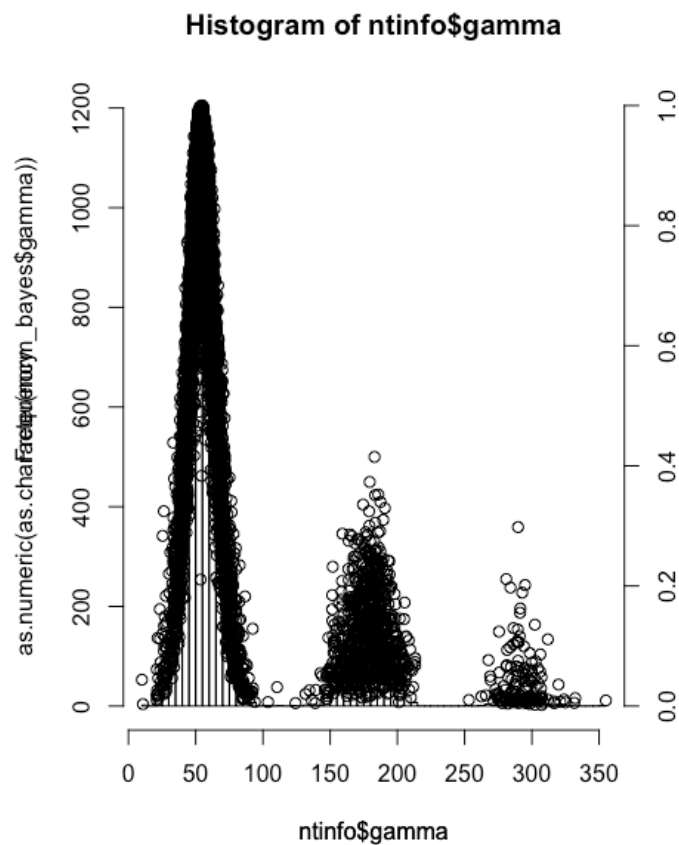
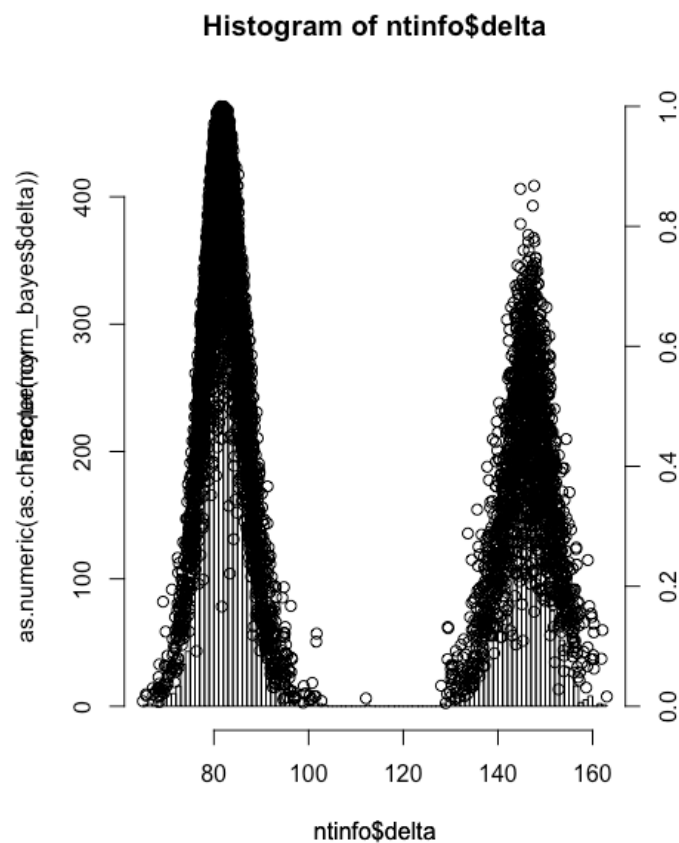
Histogram of ntinfo\$beta



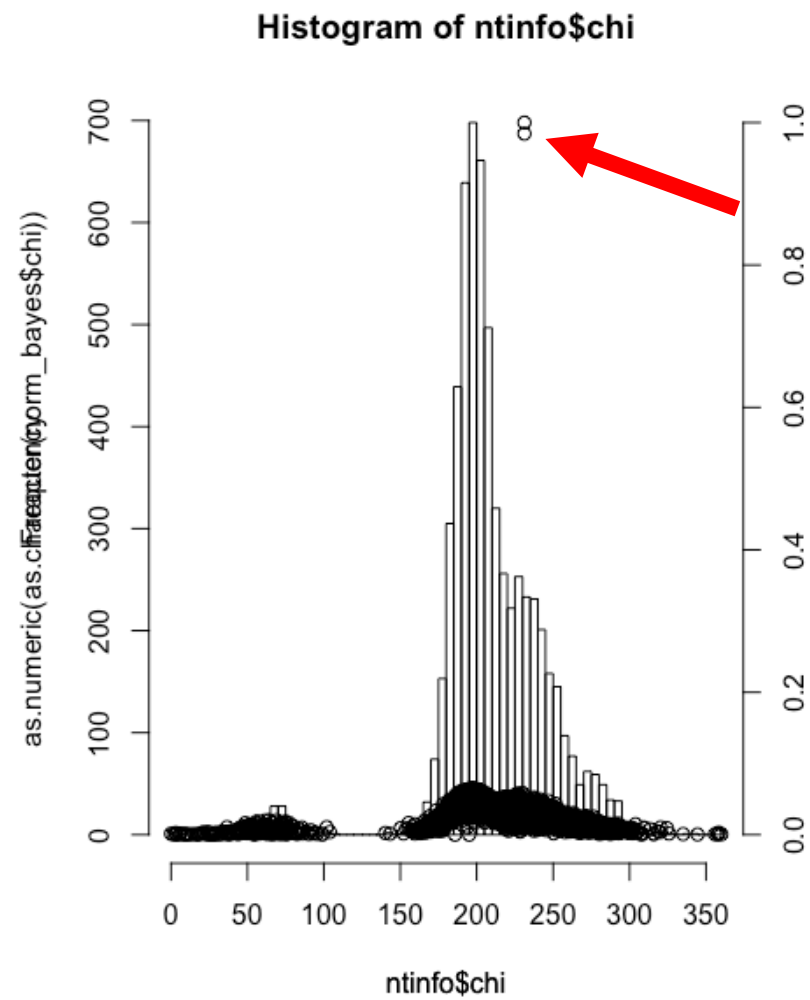
Histogram of ntinfo\$zeta



Check all the plots generated



Check all the plots generated



Stripe the highest values for beta and chi scores

•IDs for BETA:

771

5031

	alpha	beta	gamma	delta	epsilon	zeta	chi
771	289.201	186.393	64.264	144.276	274.418	85.945	225.393
5031	305.376	186.394	57.141	159.383	292.826	86.494	249.556

•Scoring by Bayes function:

	alpha	beta	gamma	delta	epsilon	zeta	chi	total
771	0.8973062	16.05909	0.5409483	0.4490677	0.47092283	0.3226892	0.6540827	2.770586
5031	0.5285648	15.77888	0.6109064	0.1482734	0.09670432	0.1870816	0.3746445	2.532151

•Scoring by cybeRNAting:

	alpha	beta	gamma	delta	epsilon	zeta	chi
1	0.922	0.262	0.441	0.015	0.024	0.012	0.079
2	0.375	0.335	0.836	0.02	0.021	0.017	0.051

•IDs for CHI:

1677

3233

	alpha	beta	gamma	delta	epsilon	zeta	chi
1677	270.223	201.464	67.534	135.958	246.685	304.819	232.028
3233	303.511	187.134	42.727	136.420	284.622	91.193	232.027

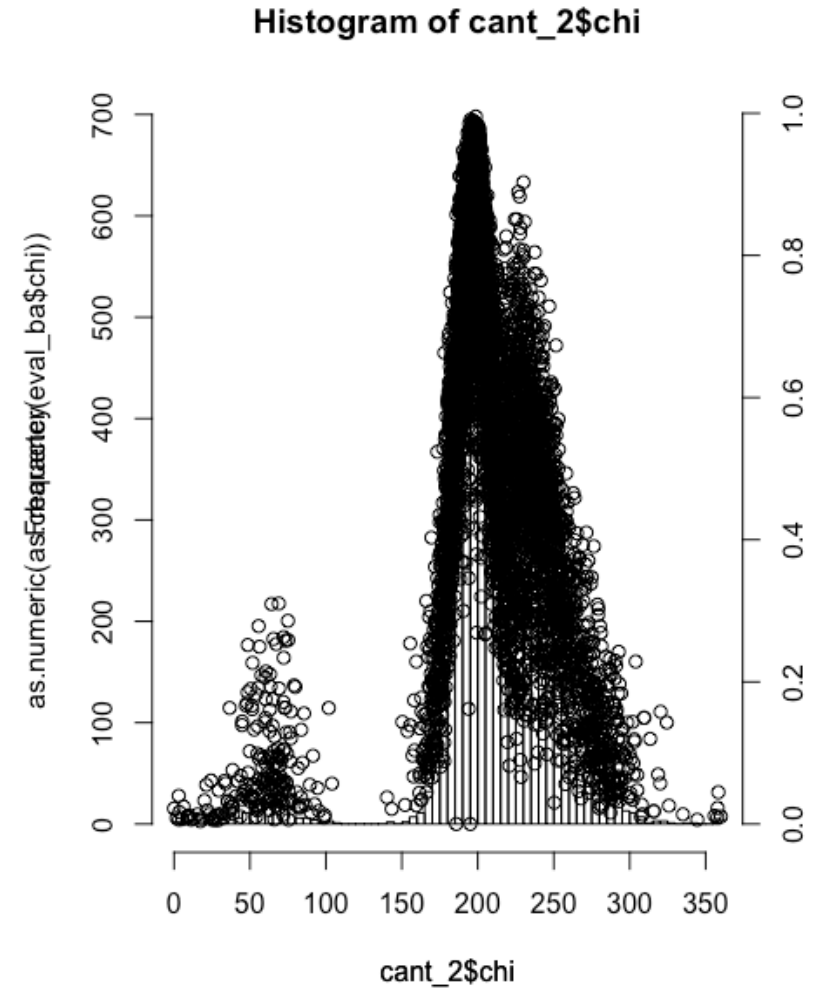
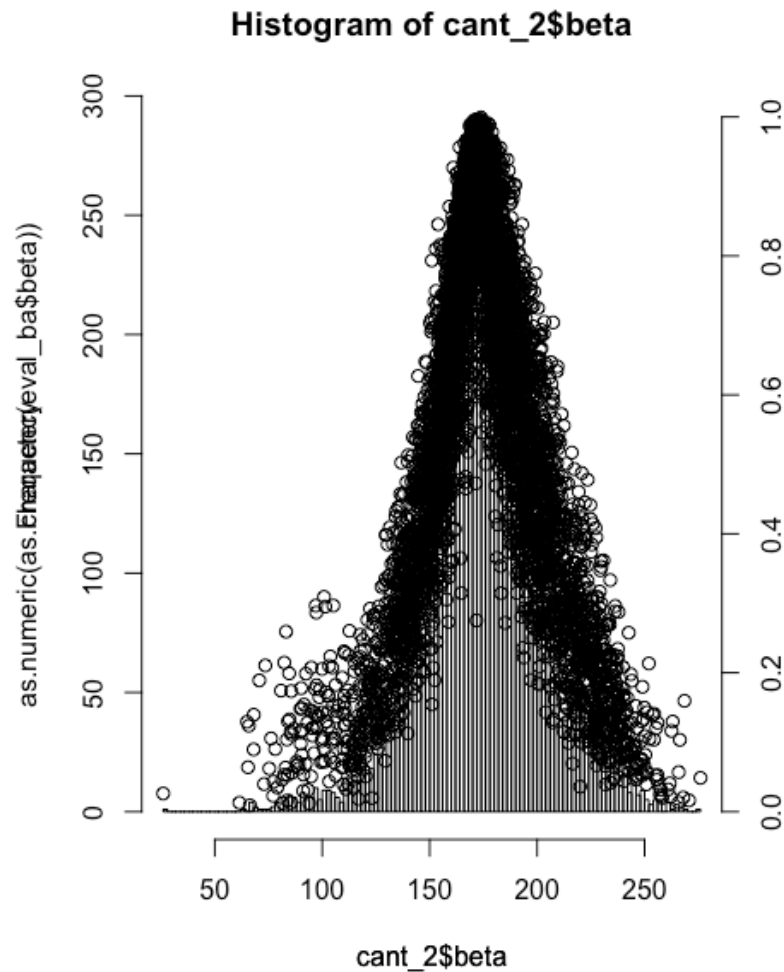
•Scoring by Bayes function:

	alpha	beta	gamma	delta	epsilon	zeta	chi	total
1677	0.3167092	0.6932020	0.5414358	0.1373346	0.8894099	0.4993746	15.17293	2.607199
3233	0.5621434	0.6215669	0.5625761	0.1392141	0.3949913	0.3781720	15.41195	2.581517


•Scoring by cybeRNAting:

	alpha	beta	gamma	delta	epsilon	zeta	chi
1	0.138	0.087	0.29	0.047	0.109	0.128	0.094
2	0.477	0.311	0.205	0.043	0.032	0.018	0.073

Remove those 4 nucleotides and check again plots for Beta and Chi



We are already normalizing with the max_dens dividing so that our data fits in the range of [0,1]



	alpha	beta	gamma	delta	epsilon	zeta	chi	total
771	0.8973062	16.05909	0.5409483	0.4490677	0.47092283	0.3226892	0.6540827	2.770586
5031	0.5285648	15.77888	0.6109064	0.1482734	0.09670432	0.1870816	0.3746445	2.532151

Does not make sense > 1 values

Deep analysis on the ID: 771. Solving its problem may solve problem for the others

	alpha	beta	gamma	delta	epsilon	zeta	chi
771	289.201	186.393	64.264	144.276	274.418	85.945	225.393

1) Scoring by Bayes function:

	alpha	beta	gamma	delta	epsilon	zeta	chi	total
771	0.8973062	16.05909	0.5409483	0.4490677	0.47092283	0.3226892	0.6540827	2.770586

2) Analysis of the different contributions “beta”

```
beta_amb <- c(nulling(dnorms_calc("alpha",eval[, "alpha"], "beta",eval2[, "beta"])),
  nulling(dnorms_calc("gamma",eval[, "gamma"], "beta",eval2[, "beta"])),
  nulling(dnorms_calc("delta",eval[, "delta"], "beta",eval2[, "beta"])),
  nulling(dnorms_calc("epsilon",eval[, "epsilon"], "beta",eval2[, "beta"])),
  nulling(dnorms_calc("zeta",eval[, "zeta"], "beta",eval2[, "beta"])),
  nulling(dnorms_calc("chi",eval[, "chi"], "beta",eval2[, "beta"])))
```

beta_amb

OUTPUT: 0.7346722 0.9257825 0.7290086 0.8479728 **92.1275127** 0.9895698

3) Check at the probability function for: zeta – beta pairs

```
#####for zeta
if(angle == "zeta"){
  zeta <- vector("list", length = length(inter_list))

  for(i in 1:length(inter_list)){
    index <- which(ntinfo_m[, "zeta"] == i)
    if(length(index) > 5){
      dataframe2 <- ntinfo_m[index, ]
      dataframe2_no_disc <- ntinfo[index, ]

      #Check for zeta analysis
      gaussians_alpha<- fit_data(na.omit(dataframe2_no_disc$alpha))
      zeta[[i]]$alpha$prop <- gaussians_alpha$prop
      zeta[[i]]$alpha$mean <- gaussians_alpha$mean
      zeta[[i]]$alpha$sd <- gaussians_alpha$sd
      zeta[[i]]$alpha$max_dens <- gaussians_alpha$max_dens

      #Check for zeta
      gaussians_beta<- fit_data(na.omit(dataframe2_no_disc$beta))
      zeta[[i]]$beta$prop <- gaussians_beta$prop
      zeta[[i]]$beta$mean <- gaussians_beta$mean
      zeta[[i]]$beta$sd <- gaussians_beta$sd
      zeta[[i]]$beta$max_dens <- gaussians_beta$max_dens
```

4) Dataframe subset for: zeta = 86

	alpha	beta	gamma	delta	epsilon	zeta	chi
150	299.985	224.918	58.309	152.590	275.147	86.087	186.036
771	289.201	186.393	64.264	144.276	274.418	85.945	225.393
1775	277.810	196.823	64.400	150.562	284.776	86.384	243.595
1969	63.340	182.985	69.275	144.326	270.723	86.313	237.007
1981	304.409	177.257	44.152	80.000	196.679	85.811	201.329
2632	270.536	178.374	69.283	85.215	278.259	85.647	200.619
3338	48.854	178.523	281.555	149.244	231.742	86.158	58.462
3446	73.771	202.162	59.702	93.546	245.278	85.703	293.130
3701	69.662	199.497	38.083	148.515	285.610	85.506	50.386
3864	297.686	179.831	50.869	148.359	242.312	86.123	241.039
4316	55.942	171.598	294.024	150.164	223.961	85.928	56.351
4490	291.336	175.557	63.874	88.631	264.995	85.927	204.186
4632	101.495	229.141	61.469	147.448	204.297	85.988	209.828
5031	305.376	186.394	57.141	159.383	292.826	86.494	249.556
5101	269.884	160.637	50.488	150.559	281.099	85.792	246.733
5994	289.182	201.943	51.546	148.065	273.845	85.614	238.402
6223	62.093	169.994	46.134	142.933	232.856	86.388	240.206

5) Fitted gaussians for beta when zeta = 86

	prop	mean	sd	max_dens
1	0.36354770	173.7021	7.27810908	0.6179944
2	0.05430858	176.5040	0.87766141	0.6179944
3	0.11202925	178.4486	0.07449993	0.6179944
4	0.11763805	186.3935	0.00050000	0.6179944
5	0.11744160	198.1713	1.34759935	0.6179944
6	0.11738776	202.0526	0.10949999	0.6179944
7	0.11764706	227.0295	2.11150000	0.6179944

6) Compute dnorm() calculus on each of them

```
> loopout(eval2 = 186.393, x = gaussians_beta, ind = 7)
[1] 0.007050696
[1] 7.050696e-03 1.080494e-29
[1] 7.050696e-03 1.080494e-29 0.000000e+00
[1] 7.050696e-03 1.080494e-29 0.000000e+00 9.212046e+01
[1] 7.050696e-03 1.080494e-29 0.000000e+00 9.212046e+01 1.452381e-18
[1] 7.050696e-03 1.080494e-29 0.000000e+00 9.212046e+01 1.452381e-18 0.000000e+00
[1] 7.050696e-03 1.080494e-29 0.000000e+00 9.212046e+01 1.452381e-18 0.000000e+00 1.343955e-82
[1] 92.12751
```

4th Gaussian gives such bad result

7) Compute dnorm() manually on the 4th row

```
> dnorm(x = 186.393, mean = 186.3935, sd = 0.00050000)
[1] 483.9414
> dnorm(x = 186.393, mean = 186.3935, sd = 0.00050000) * 0.11763805
[1] 56.92993
> dnorm(x = 186.393, mean = 186.3935, sd = 0.00050000) * 0.11763805 / 0.6179944
[1] 92.12046
```


PROBLEM:

·Standard deviation is too low so minimal changes on x versus mean create a lot of change at dnorm()

SOLUTION:

·Apply filtering for sd
·Data handling

***Trying with different values on x:

```
> dnorm(x = 184, mean = 186.3935, sd = 0.00050000) * 0.11763805 / 0.6179944
[1] 0
> dnorm(x = 186, mean = 186.3935, sd = 0.00050000) * 0.11763805 / 0.6179944
[1] 0
> dnorm(x = 187, mean = 186.3935, sd = 0.00050000) * 0.11763805 / 0.6179944
[1] 0
> dnorm(x = 186.37, mean = 186.3935, sd = 0.00050000) * 0.11763805 / 0.6179944
[1] 0
> dnorm(x = 186.395, mean = 186.3935, sd = 0.00050000) * 0.11763805 / 0.6179944
[1] 1.687245
> dnorm(x = 186.3933, mean = 186.3935, sd = 0.00050000) * 0.11763805 / 0.6179944
[1] 140.2038
```