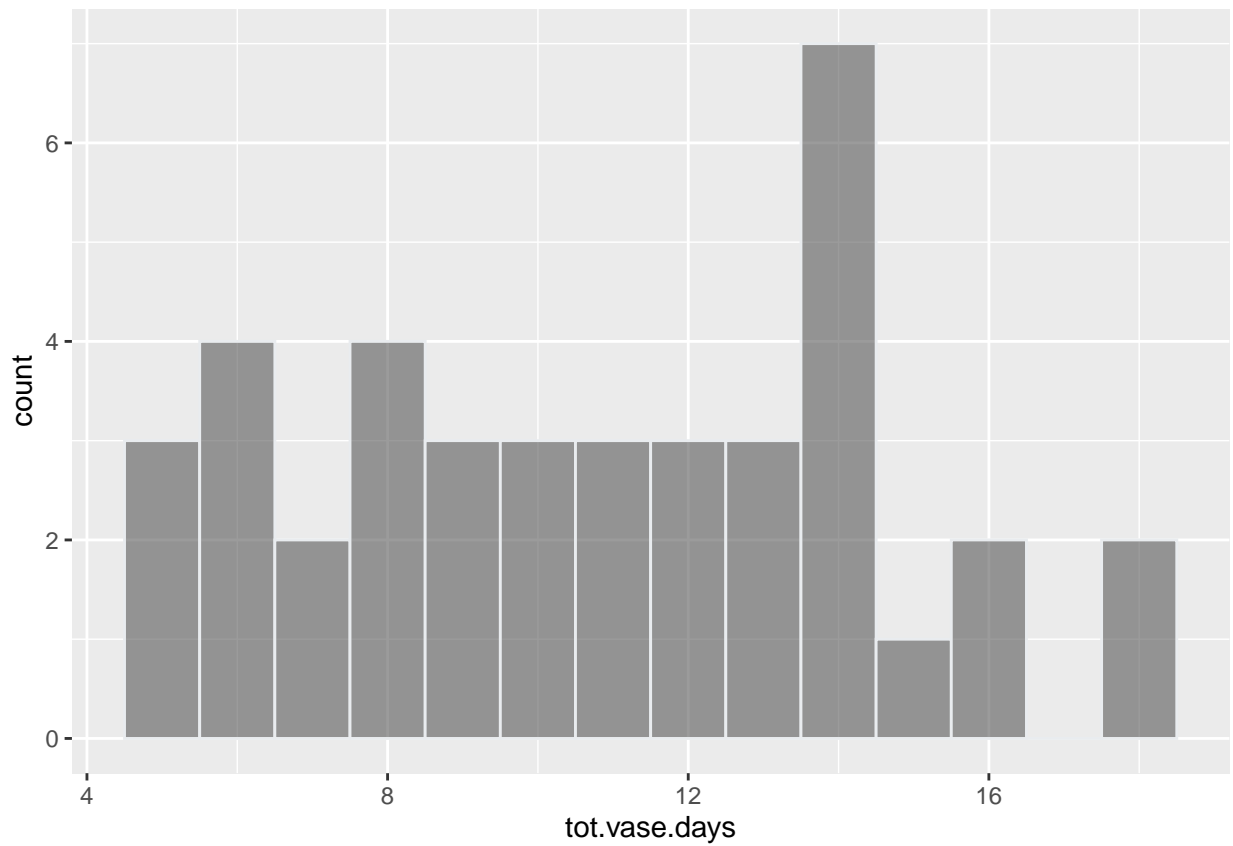


Analysis of the pilot data - PMHD - Group 6

The data

The data is loaded and visualized with a histogram.

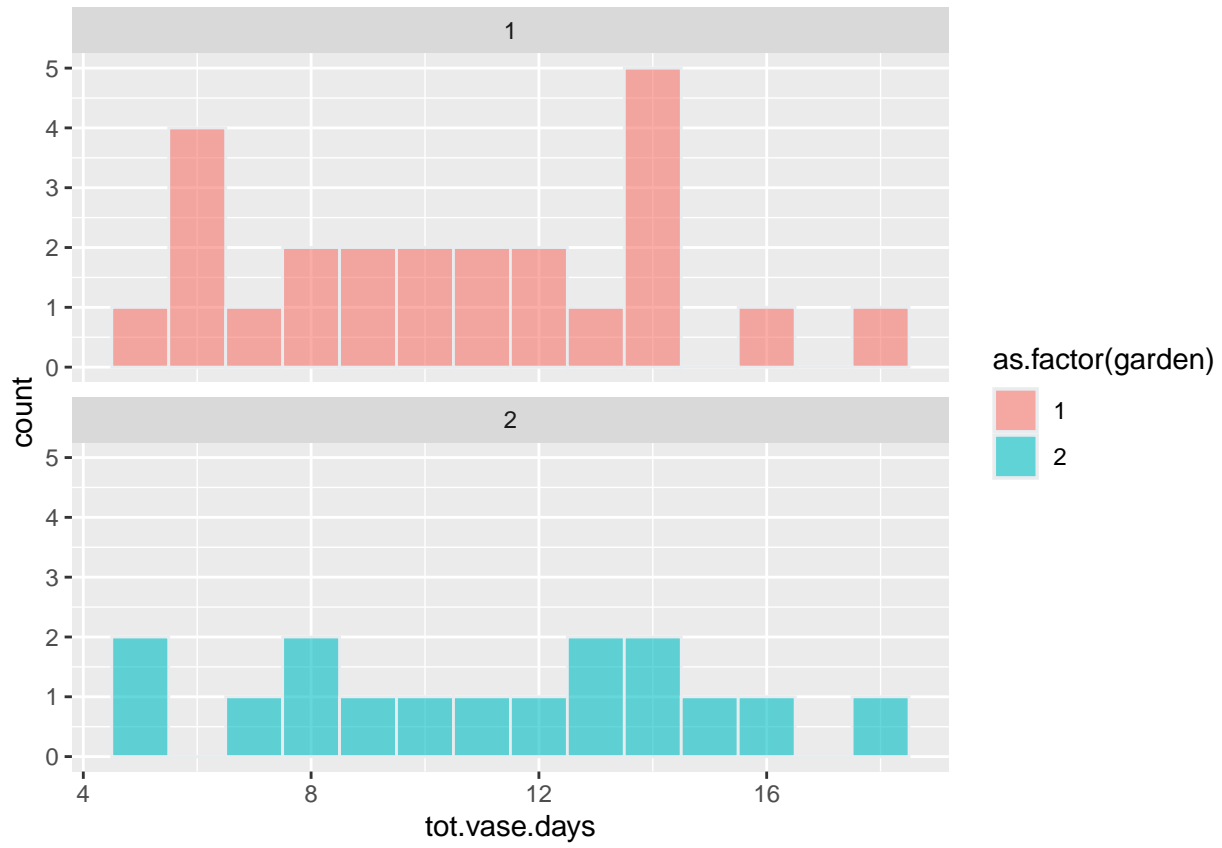
```
pilot.data <- read.csv("G6.pilot.data.csv")
ggplot(data=pilot.data, aes(x=tot.vase.days)) +
  geom_histogram(color="#e9ecef", alpha=0.6, position = 'identity', binwidth = 1)
```



Plotting the histograms based on the different possible sources of variability

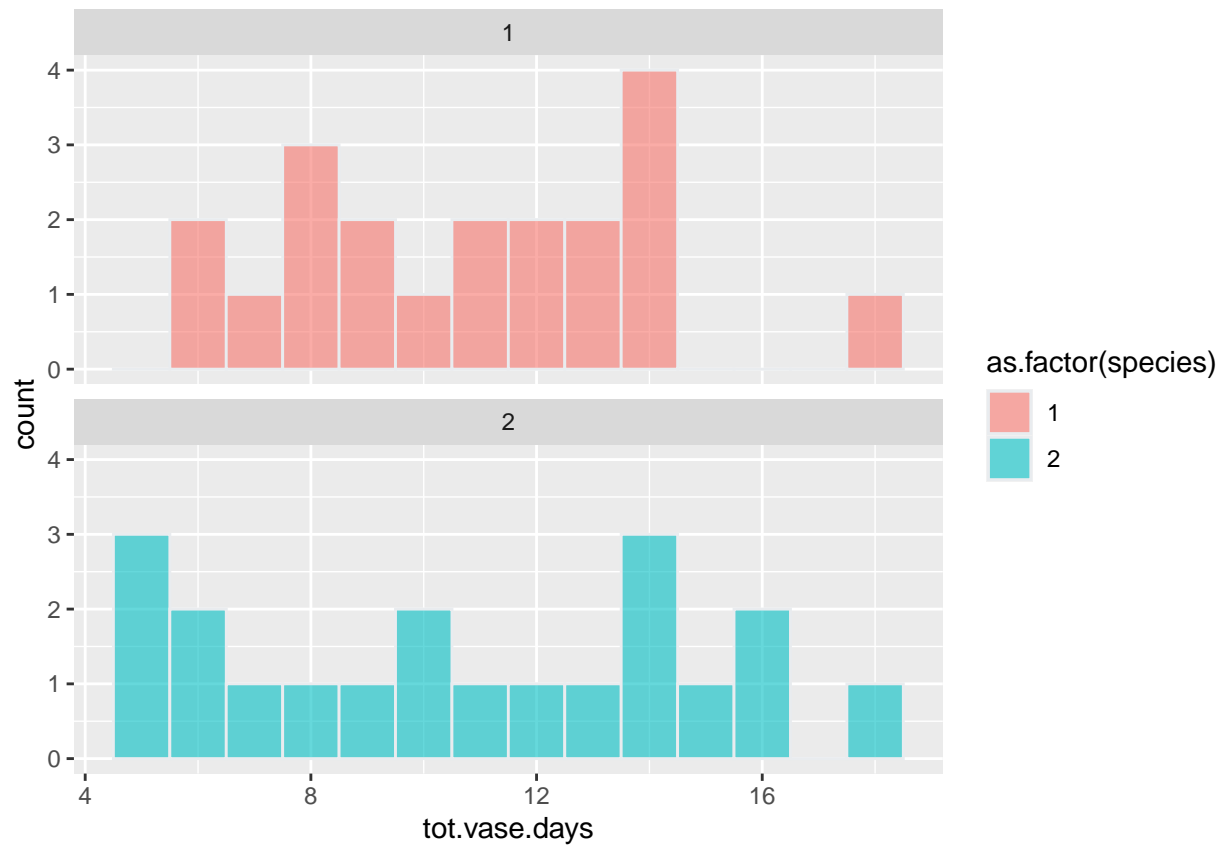
Garden

```
ggplot(data=pilot.data, aes(x=tot.vase.days, fill=as.factor(garden))) +  
  geom_histogram( color="#e9ecef", alpha=0.6, position = 'identity', binwidth = 1) +  
  facet_wrap(~as.factor(garden),ncol=1)
```



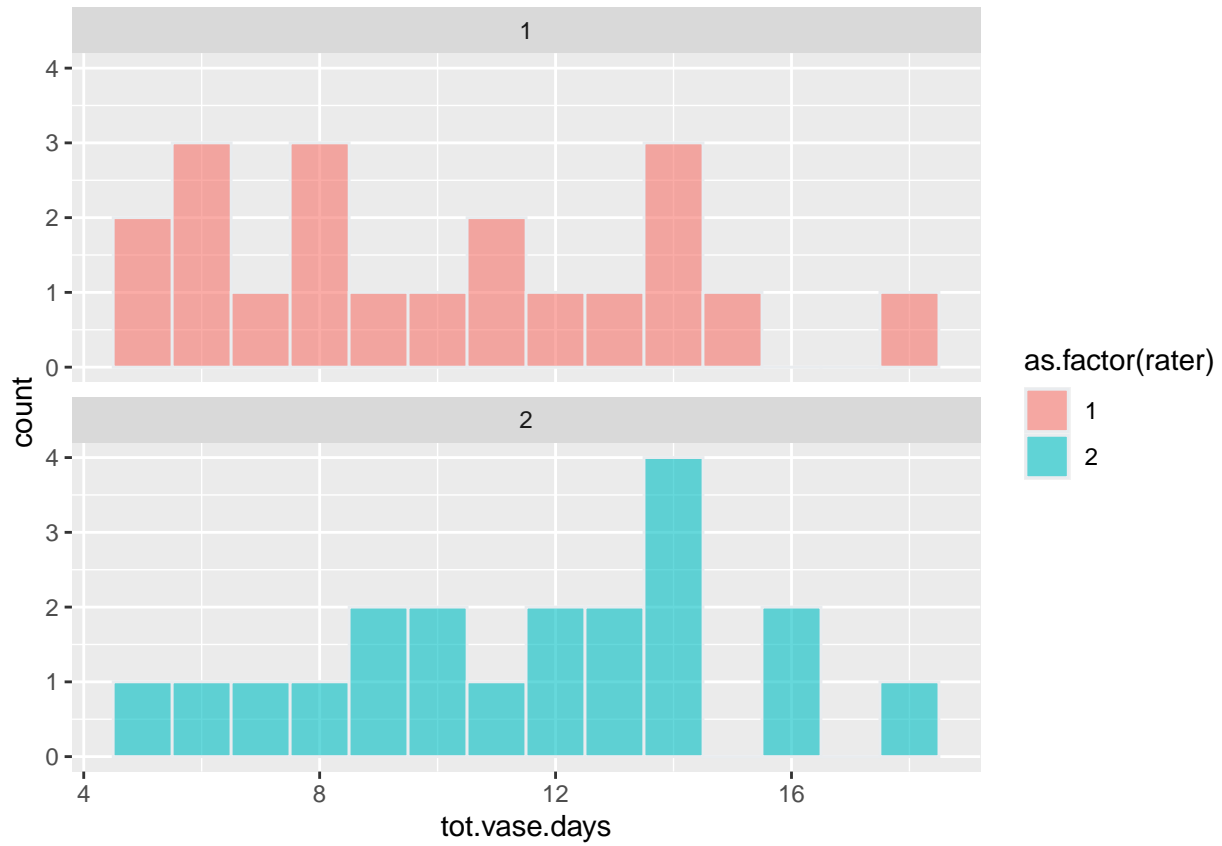
Species

```
ggplot(data=pilot.data, aes(x=tot.vase.days, fill=as.factor(species))) +  
  geom_histogram( color="#e9ecef", alpha=0.6, position = 'identity', binwidth = 1) +  
  facet_wrap(~as.factor(species),ncol=1)
```



Rater

```
ggplot(data=pilot.data, aes(x=tot.vase.days, fill=as.factor(rater))) +  
  geom_histogram( color="#e9ecef", alpha=0.6, position = 'identity', binwidth = 1) +  
  facet_wrap(~as.factor(rater),ncol=1)
```



Estimation of the lambda parameter and confidence interval

The lambda parameter is estimated by fitting the pilot data to a poisson distribution.

```
parms <- fitdistr(pilot.data$tot.vase.days, "poisson")
lambda.pilot <- parms$estimate
sd_x <- parms$sd
confidence.interval.95 <- c(lambda.pilot + c(-1,1) * 1.96 * sd_x)
lambda.pilot
```

```
## lambda
## 10.775
```

```
confidence.interval.95
```

```
## [1] 9.757734 11.792266
```

Simulation to calculate the sample size

Simulation based on the code provided by the professor in the lecture of sample size calculation.

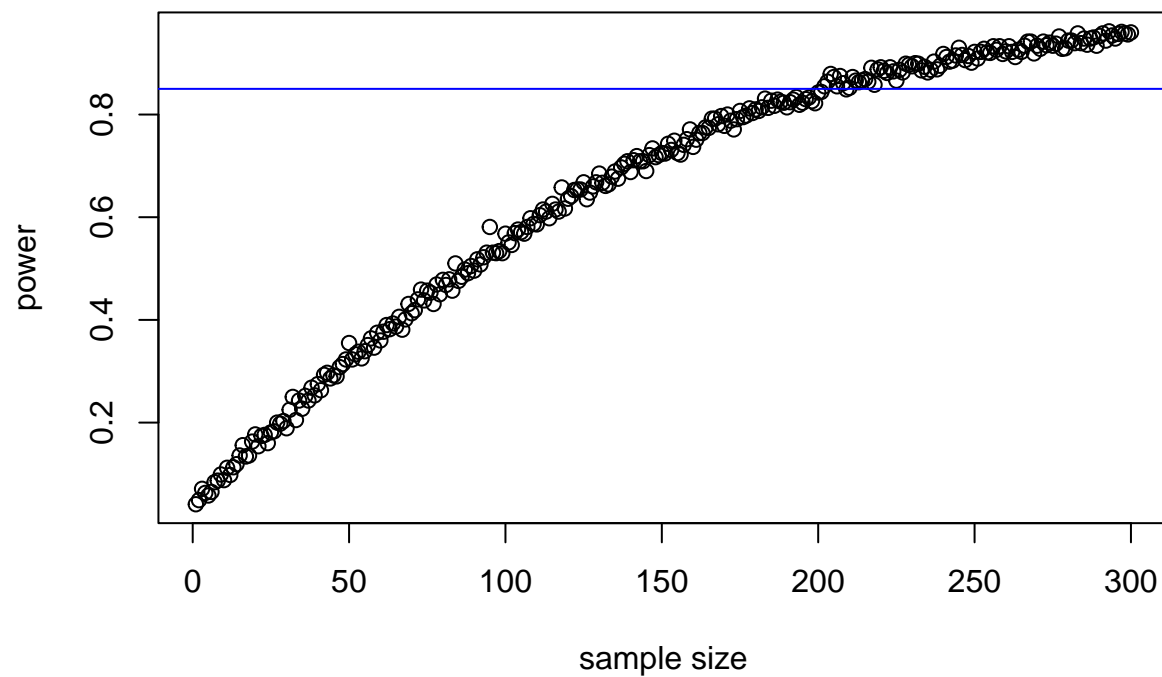
```
max.Sample.Size <- 300 # we evaluate all possible sample sizes up to 300
power <- rep(0,max.Sample.Size) # initialization of the variable
for (N in 1:max.Sample.Size){
  # setting the lambda parameters of the two populations
  lambda.control = lambda.pilot # this value may come from pilot study
  lambda.treated = lambda.pilot+1 # value of the control group plus the effect size
  alpha = 0.05 # significance level

  # simulation of a high number of experiments and test
  numberSimulation <- 1000
  pval <- numeric(numberSimulation) # initialization of the variable

  set.seed(1234) # set the seed for reproducible results
  for (i in 1:numberSimulation){
    # we simulate from Poisson distribution
    controlGroup <- rpois(N, lambda = lambda.control)
    treatedGroup <- rpois(N, lambda = lambda.treated)
    simData <- data.frame(response = c(controlGroup, treatedGroup), treatment = rep(c(0, 1), each = N))
    # we use a GLM model for Poisson regression to test effect of treatment
    pval[i] <- summary(glm(response ~ treatment, data = simData, family=poisson()))$coeff["treatment"]
  }

  # Estimation of the power
  power[N]<-sum(pval < alpha)/numberSimulation
}

# plotting the results
sample.size <- 1:max.Sample.Size
plot(sample.size, power,xlab = "sample size")
abline(h=0.85, col="blue")
```



```
results <- data.frame(sample.size,power)
results
```

##	sample.size	power
## 1	1	0.041
## 2	2	0.049
## 3	3	0.071
## 4	4	0.063
## 5	5	0.058
## 6	6	0.065
## 7	7	0.084
## 8	8	0.087
## 9	9	0.099
## 10	10	0.088
## 11	11	0.112
## 12	12	0.098
## 13	13	0.113
## 14	14	0.119
## 15	15	0.136
## 16	16	0.156
## 17	17	0.134
## 18	18	0.136
## 19	19	0.163
## 20	20	0.177
## 21	21	0.154
## 22	22	0.174

## 23	23 0.176
## 24	24 0.160
## 25	25 0.181
## 26	26 0.183
## 27	27 0.200
## 28	28 0.198
## 29	29 0.203
## 30	30 0.189
## 31	31 0.225
## 32	32 0.250
## 33	33 0.205
## 34	34 0.243
## 35	35 0.227
## 36	36 0.252
## 37	37 0.243
## 38	38 0.268
## 39	39 0.253
## 40	40 0.275
## 41	41 0.263
## 42	42 0.293
## 43	43 0.297
## 44	44 0.286
## 45	45 0.292
## 46	46 0.290
## 47	47 0.308
## 48	48 0.314
## 49	49 0.323
## 50	50 0.355
## 51	51 0.323
## 52	52 0.333
## 53	53 0.338
## 54	54 0.325
## 55	55 0.339
## 56	56 0.351
## 57	57 0.364
## 58	58 0.346
## 59	59 0.375
## 60	60 0.360
## 61	61 0.377
## 62	62 0.390
## 63	63 0.382
## 64	64 0.393
## 65	65 0.387
## 66	66 0.406
## 67	67 0.381
## 68	68 0.401
## 69	69 0.431
## 70	70 0.413
## 71	71 0.419
## 72	72 0.440
## 73	73 0.459
## 74	74 0.438
## 75	75 0.457
## 76	76 0.453

## 77	77 0.431
## 78	78 0.469
## 79	79 0.450
## 80	80 0.478
## 81	81 0.468
## 82	82 0.479
## 83	83 0.457
## 84	84 0.510
## 85	85 0.476
## 86	86 0.484
## 87	87 0.498
## 88	88 0.491
## 89	89 0.505
## 90	90 0.496
## 91	91 0.518
## 92	92 0.508
## 93	93 0.522
## 94	94 0.531
## 95	95 0.581
## 96	96 0.531
## 97	97 0.530
## 98	98 0.534
## 99	99 0.530
## 100	100 0.568
## 101	101 0.551
## 102	102 0.546
## 103	103 0.569
## 104	104 0.576
## 105	105 0.571
## 106	106 0.568
## 107	107 0.581
## 108	108 0.598
## 109	109 0.587
## 110	110 0.586
## 111	111 0.604
## 112	112 0.615
## 113	113 0.611
## 114	114 0.598
## 115	115 0.626
## 116	116 0.615
## 117	117 0.611
## 118	118 0.658
## 119	119 0.617
## 120	120 0.636
## 121	121 0.641
## 122	122 0.653
## 123	123 0.654
## 124	124 0.654
## 125	125 0.668
## 126	126 0.635
## 127	127 0.648
## 128	128 0.660
## 129	129 0.668
## 130	130 0.685

## 131	131 0.667
## 132	132 0.661
## 133	133 0.664
## 134	134 0.679
## 135	135 0.689
## 136	136 0.675
## 137	137 0.697
## 138	138 0.704
## 139	139 0.709
## 140	140 0.688
## 141	141 0.711
## 142	142 0.719
## 143	143 0.709
## 144	144 0.709
## 145	145 0.690
## 146	146 0.721
## 147	147 0.734
## 148	148 0.717
## 149	149 0.721
## 150	150 0.725
## 151	151 0.724
## 152	152 0.743
## 153	153 0.731
## 154	154 0.749
## 155	155 0.725
## 156	156 0.722
## 157	157 0.741
## 158	158 0.752
## 159	159 0.771
## 160	160 0.737
## 161	161 0.751
## 162	162 0.764
## 163	163 0.764
## 164	164 0.775
## 165	165 0.774
## 166	166 0.792
## 167	167 0.792
## 168	168 0.781
## 169	169 0.797
## 170	170 0.778
## 171	171 0.800
## 172	172 0.788
## 173	173 0.771
## 174	174 0.791
## 175	175 0.807
## 176	176 0.795
## 177	177 0.798
## 178	178 0.812
## 179	179 0.803
## 180	180 0.810
## 181	181 0.807
## 182	182 0.814
## 183	183 0.831
## 184	184 0.813

## 185	185 0.827
## 186	186 0.817
## 187	187 0.829
## 188	188 0.825
## 189	189 0.823
## 190	190 0.814
## 191	191 0.824
## 192	192 0.829
## 193	193 0.834
## 194	194 0.819
## 195	195 0.823
## 196	196 0.830
## 197	197 0.833
## 198	198 0.826
## 199	199 0.822
## 200	200 0.843
## 201	201 0.844
## 202	202 0.855
## 203	203 0.864
## 204	204 0.879
## 205	205 0.873
## 206	206 0.855
## 207	207 0.875
## 208	208 0.861
## 209	209 0.849
## 210	210 0.852
## 211	211 0.873
## 212	212 0.866
## 213	213 0.863
## 214	214 0.866
## 215	215 0.869
## 216	216 0.864
## 217	217 0.891
## 218	218 0.858
## 219	219 0.888
## 220	220 0.892
## 221	221 0.885
## 222	222 0.881
## 223	223 0.892
## 224	224 0.884
## 225	225 0.866
## 226	226 0.887
## 227	227 0.882
## 228	228 0.899
## 229	229 0.897
## 230	230 0.894
## 231	231 0.900
## 232	232 0.899
## 233	233 0.886
## 234	234 0.893
## 235	235 0.882
## 236	236 0.887
## 237	237 0.903
## 238	238 0.888

## 239	239 0.895
## 240	240 0.918
## 241	241 0.912
## 242	242 0.903
## 243	243 0.905
## 244	244 0.915
## 245	245 0.930
## 246	246 0.916
## 247	247 0.906
## 248	248 0.914
## 249	249 0.901
## 250	250 0.922
## 251	251 0.909
## 252	252 0.922
## 253	253 0.928
## 254	254 0.921
## 255	255 0.920
## 256	256 0.933
## 257	257 0.926
## 258	258 0.933
## 259	259 0.918
## 260	260 0.923
## 261	261 0.933
## 262	262 0.922
## 263	263 0.912
## 264	264 0.926
## 265	265 0.922
## 266	266 0.934
## 267	267 0.942
## 268	268 0.942
## 269	269 0.919
## 270	270 0.934
## 271	271 0.928
## 272	272 0.942
## 273	273 0.935
## 274	274 0.940
## 275	275 0.934
## 276	276 0.938
## 277	277 0.952
## 278	278 0.928
## 279	279 0.929
## 280	280 0.944
## 281	281 0.944
## 282	282 0.939
## 283	283 0.958
## 284	284 0.939
## 285	285 0.948
## 286	286 0.936
## 287	287 0.948
## 288	288 0.950
## 289	289 0.934
## 290	290 0.953
## 291	291 0.958
## 292	292 0.944

## 293	293 0.962
## 294	294 0.954
## 295	295 0.948
## 296	296 0.957
## 297	297 0.961
## 298	298 0.958
## 299	299 0.956
## 300	300 0.960