Report2

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1 Report 2

Authors: Richard Liang, Olivia majedi, Priyanka Kishore, Rhea Rakheja, Jiadong Li, Michael Strobel

```
[1]: import pandas as pd
[2]: df = pd.read_csv('https://raw.githubusercontent.com/jdli28/STAT440/master/
      ⇔summer_mount_ginini.csv')
[3]: df = df.dropna()
[4]: df
[4]:
               Date
                        Location MinTemp
                                            MaxTemp
     0
          12/1/2008
                     MountGinini
                                       5.2
                                               13.0
     1
          12/2/2008
                     MountGinini
                                       3.0
                                               15.0
     2
          12/3/2008
                     MountGinini
                                       6.0
                                               15.0
     3
          12/4/2008
                     MountGinini
                                       2.0
                                               15.0
     4
                                               17.8
          12/5/2008
                     MountGinini
                                       8.0
     . .
     743
         2/24/2017
                     MountGinini
                                      13.8
                                               24.1
         2/25/2017
                                       9.5
                                               11.2
     744
                     MountGinini
     745
         2/26/2017
                     MountGinini
                                       4.7
                                               16.7
     746 2/27/2017
                     MountGinini
                                       5.6
                                               16.2
     747 2/28/2017 MountGinini
                                       7.5
                                               18.0
     [745 rows x 4 columns]
[5]: import numpy as np
```

1.1 Population size N and true parameter mu(MaxTemp)

```
[6]: N = len(df)
N
```

```
[6]: 745
```

```
[7]: max_temp_mean = np.mean(df.MaxTemp)
max_temp_mean
```

[7]: 19.112885906040272

1.2 Calculate sample size n for 90% and 95% confidence levels and couple different d's. Use true ^2 for these calculations

```
[8]: sigma_sq = np.var(df.MaxTemp)
      sigma_sq
 [8]: 21.75814267825774
 [9]: r = [.05, .01, .1]
      d = [(max_temp_mean * rval) for rval in r]
      d
 [9]: [0.9556442953020137, 0.19112885906040272, 1.9112885906040273]
[10]: from scipy import stats
      z_{alpha_90} = stats.norm.ppf(1-0.05)
      z_{alpha_95} = stats.norm.ppf(1-0.025)
[11]: n_90 = []
      n_95 = []
[12]: for d_val in d:
        n0 = z_{alpha_90**2*sigma_sq/(d_val**2)}
        n_90.append(
          1/((1/n0)+(1/N))
        )
        n0 = z_{alpha_95**2*sigma_sq/(d_val**2)}
        n_95.append(
          1/((1/n0)+(1/N))
        )
```

[13]: import math

```
[14]: n_90
n_90 = [math.ceil(n) for n in n_90]
n_90
```

[14]: [60, 510, 16]

```
[15]: n_95
n_95 = [math.ceil(n) for n in n_95]
n_95
```

[15]: [82, 563, 23]

1.3 Estimate your parameter of interest using SRS with n's which you calculated above.

1.3.1 90% CI

```
[16]: sample_90s = [] sample_95s = []
```

```
[17]: for n in n_90:
    sample = np.random.choice(df.MaxTemp, size=n, replace=False)
    sample_90s.append(sample)
    print(np.mean(sample))
```

- 19.108333333333334
- 19.12117647058824
- 20.25625

1.3.2 95% CI

```
[18]: for n in n_95:
    sample = np.random.choice(df.MaxTemp, size=n, replace=False)
    sample_95s.append(sample)
    print(np.mean(sample))
```

- 19.954878048780486
- 19.223445825932505
- 17.352173913043476

1.4 Estimate variance of your estimator for these n's

```
[19]: for sample in sample_90s:
    print(np.var(sample))
```

- 17.410763888888887
- 22.90453194925029
- 9.679960937499999

```
[20]: for sample in sample_95s:
    print(np.var(sample))

28.51686644854253
21.62307373907227
18.305973534971645
```

1.5 Calculate confidence intervals for these estimators.

1.5.1 90% CI for ybar

```
for sample in sample_90s:
    ybar = np.mean(sample)
    n = len(sample)
    s_sq = np.var(sample)
    ci = get_ybar_CI(ybar, z_alpha_90, n, s_sq)
    print(ci)
```

```
[18.25870752919851, 19.95795913746816]
[18.925400820732197, 19.31695212044428]
[18.990669730177896, 21.521830269822107]
```

1.5.2 95% CI for ybar

```
[23]: for sample in sample_95s:
    ybar = np.mean(sample)
    n = len(sample)
    s_sq = np.var(sample)
    ci = get_ybar_CI(ybar, z_alpha_95, n, s_sq)
    print(ci)
```

```
[18.864516084072232, 21.04524001348874]
[19.033596016347715, 19.413295635517294]
[15.630816514537186, 19.073531311549768]
```

1.6 Choosing optimal sample sizes

```
[24]: n = n_95[0]
n
```

[24]: 82

1.7 Guaranteeing the nominal confidence level

```
[25]: d_val = 0.9556442953020137
[26]: ct = 0
      for i in range(100):
        sample = np.random.choice(df.MaxTemp, size=n, replace=False)
        ybar = np.mean(sample)
        d = abs(ybar - max_temp_mean)
        print(d)
        if d - d_val < 0:</pre>
          ct +=1
      print("----")
      print(ct)
     0.4970322475036859
     0.7702029792110068
     0.22491897200850985
     0.06882141103289996
     0.21760189883777414
     0.3677639548207594
     0.5188214110328992
     0.0700409232280208
     0.5663823866426512
     0.13849566213783504
     0.10784580127679888
     0.5287395645768562
     0.2628859060402746
     0.39337371091832196
     0.08833360615484764
     1.4968701915207028
     0.2494712718939276
     0.08711409395973035
     0.7458127353085651
     0.572479947618266
     0.09581273530856649
     0.35556883286954033
     0.29443116713045825
     0.0029677524963105384
```

- 0.12613848420363283
- 0.990772630545095
- 0.43101653298412046
- 0.05434932067441878
- 0.17508102799149228
- 0.8078458012768053
- 0.18239810116222444
- 1.0458127353085658
- 0.6823981011622209
- 0.6224799476182632
- 0.11532493043051417
- 0.042154198723196146
- 0.16516287444753175
- 0.7493092159109445
- 0.3517482403011911
- 0.5958127353085665
- 0.5006907840890555
- 0.757845801276801
- 0.1993092159109473
- 0.45191029628417567
- 0.2761384842036314
- 0.8066262890816809
- 0.11882141103289712
- 0.2970322475036866
- 0.4884956621378329
- 0.1749189720085056
- 0.37873956457685765
- 0.009227369454904988
- 0.28101653298411833
- 0.36776395482076296
- 1.1651628744475317
- 0.28239810116222586
- 0.1726420036012506
- 0.34947127189393257
- 0.18605663774759051
- 0.7567883450646633
- 0.24808970371582717
- 0.028739564576856225
- 0.5066262890816766
- 0.2665444426256407
- 0.11882141103289712
- 0.14337371091832196
- 0.9310165329841205
- 0.5555688328695396
- 0.15312980847929936
- 0.6702029792110018
- 0.4531298084792965
- 0.1445932231134428

- 0.02264200360124846
- 0.8836176133573446
- 0.507845801276801
- 0.45922736945490783
- 0.43711409395972467
- 0.5005287281060653
- 0.6102848256670441
- 0.37264200360124633
- 0.5458127353085658
- 0.6310165329841162
- 1.018983467015886
- 0.594431167130459
- 0.012885906040271067
- 0.1895531183499699
- 0.8689834670158838
- 0.3031298084792944
- 0.07630054018661525
- 0.6176018988377727
- 0.14703224750368804
- 0.1310165329841162
- 0.16638238664264904
- 0.03727614994271278
- 0.14947127189392972
- 0.46150433786216993
- 0.4013043376021099
- 0.4456506793255848
- 0.21654444262564
- 0.03849566213783362
- 0.3651628744475346

95

Exactly 95/100 of our samples have differences less than d=0.9556442953020137

[]: