Evaluation Tasks and Description

A description on what evaluation tasks you have done and the purpose of the tasks.

Data Analysis

Data used for analysis was https://www.kaggle.com/osmi/mental-health-in-tech-survey.
Data analysis was done in R to identify the possible privacy attacks. Two different scenarios were introduced to better the consequences of attacks. Data description and analysis scenarios are described in datasource.pdf.

Implementation:

The implementation is done to demonstrate effectiveness of RAPPOR technique in preserving the privacy of individual data from the survey. The first step towards to implementation is to convert strings to bit values using hashing techniques. These bit values represent data from an individual. Bloom filters are applied to the user inputs which is followed RAPPOR encoding. This encoded data and parameters such as h, k m, p, q, f, used for encoding are sent to server.

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]	[,11]	[,12]	[,13]	[,14]	[,15]	[,16]	
30to40MaleUSNo	1	14	19	29	45	46	54	57	74	77	91	96	97	108	116	128	
30to40MaleUSYes	3	7	24	28	42	46	49	61	68	70	88	96	107	112	113	115	
40to50MaleUSYes	1	10	30	31	36	43	49	55	68	70	91	93	104	121	NA	NA	
40to50MaleUSNo	8	10	29	31	38	44	59	61	74	80	94	96	107	112	116	122	
20to30MaleUSNo	2	5	24	28	34	53	58	71	81	95	98	100	114	122	NA	NA	
20to30MaleUSYes	4	6	31	38	47	55	58	67	72	81	96	98	105	119	122	NA	
30to40MaleUKNo	4	7	25	26	42	48	54	60	74	77	86	90	98	104	119	123	
30to40MaleUKYes	9	25	28	33	36	50	59	67	78	88	95	102	111	119	128	NA	
40to50MaleUKYes	4	10	21	25	35	46	52	60	73	76	83	88	99	104	116	127	
40to50MaleUKNo	1	12	17	19	33	43	50	59	66	73	86	94	102	104	126	127	
20to30MaleUKNo	2	9	25	26	34	37	53	61	75	76	83	93	105	109	113	124	
20to30MaleUKYes	4	8	25	30	39	43	57	58	72	80	94	96	98	117	NA	NA	
30to40MaleCanadaNo	7	16	20	23	33	37	51	53	65	77	82	88	105	107	114	NA	
30to40MaleCanadaYes	6	8	23	29	39	48	58	60	71	77	86	87	103	105	116	125	
40to50MaleCanadaYes	5	7	21	26	33	36	53	54	75	80	87	95	102	109	122	125	
40to50CanadaUKNo	2	4	24	31	45	54	56	67	72	83	91	100	111	114	123	NA	
20to30MaleCanadaNo	3	10	17	28	42	43	53	60	68	75	87	89	97	105	113	124	
20to30MaleCanadaYes	8	16	23	24	37	45	57	62	76	79	82	86	104	108	117	122	

Fig. 1 Bloom filter bits vs. input – Bloom filter is represented by 16 bits (k). It means each row of the survey is represented as 16 bits.

```
> params_4x2 <- list(k = 16, m = 8, h = 2,p=0.5,q=0.75,f=0.5)
```

- > popparams=list(18,1,"Linear",0,0.05)
- > GenerateSamples(10000,params_4x2,popparams)

Fig. 2 Encoding input data

GenerateSamples method mentioned in the screenshot (Fig. 2) generates 10000 samples for user values from survey and perform encoding on the data. Results of RAPPOR encoding is given in below screenshot (Fig.3). It is a matrix of 17 candidate strings and 8 cohorts. Each value in the matrix represents rappor encoded values for each candidate string.

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]	[,11]	[,12]	[,13]	[,14]	[,15]	[,16]	[,17]
[1,]	1255	754	730	730	768	716	736	745	726	728	719	675	720	682	727	671	728
[2,]	1218	661	670	711	677	694	682	721	723	735	725	668	740	702	699	735	666
[3,]	1271	732	718	722	732	761	728	723	728	707	752	752	719	718	745	715	746
[4,]	1280	768	730	730	705	758	753	748	699	765	752	712	762	756	725	720	716
[5,]	1251	706	719	700	764	704	721	702	707	706	734	706	764	726	693	703	743
[6,]	1329	771	747	775	784	754	801	747	771	751	796	815	746	772	778	762	817
[7,]	1219	682	738	686	722	672	683	675	735	704	696	717	676	704	684	725	741
[8,]	1177	698	668	696	694	683	654	690	639	676	719	680	660	672	691	683	683

Fig. 3 RAPPOR Encoded data

This technique can preserve data privacy by sending the encoded string to server and not the actual data. Now that all the values are encoded, it is followed by decode function which could be used for publishing and for further statistical analysis.

RAPPOR encoded values, Bloom filter map (Fig. 1), parameters (Fig. 2) are sent to the decode function. The obtained results are given in the below screen shots

```
string estimate std_error proportion prop_std_error prop_low_95 prop_high_95 Truth
30to40MaleUKNo
                         30to40MaleUKNo
                                            1432
                                                       255
                                                               0.1432
                                                                              0.0255
                                                                                        0.093220
                                                                                                      0.193180
30to40MaleUSYes
                        30to40MaleUSYes
                                            1325
                                                       291
                                                               0.1325
                                                                              0.0291
                                                                                        0.075464
                                                                                                      0.189536
                                                                                                                1016
40to50MaleUSYes
                        40to50MaleUSYes
                                            1108
                                                       174
                                                               0.1108
                                                                              0.0174
                                                                                        0.076696
                                                                                                      0.144904
                                                                                                                 888
20to30MaleUSNo
                         20to30MaleUSNo
                                             857
                                                       243
                                                               0.0857
                                                                              0.0243
                                                                                        0.038072
                                                                                                      0.133328
                                                                                                                 814
40to50MaleUSNo
                         40to50MaleUSNo
                                             747
                                                       275
                                                               0.0747
                                                                              0.0275
                                                                                        0.020800
                                                                                                      0.128600
                                                                                                                 892
20to30MaleCanadaYes 20to30MaleCanadaYes
                                             715
                                                       177
                                                               0.0715
                                                                              0.0177
                                                                                        0.036808
                                                                                                      0.106192
                                                                                                                  64
20to30MaleUKYes
                        20to30MaleUKYes
                                             684
                                                       319
                                                               0.0684
                                                                              0.0319
                                                                                        0.005876
                                                                                                      0.130924
                                                                                                                 415
20to30MaleUSYes
                        20to30MaleUSYes
                                                       247
                                                               0.0615
                                                                               0.0247
                                                                                        0.013088
                                                                                                      0.109912
$summary
                   parameters
                                    values
            Candidate strings
                                   18.0000
2
            Detected strings
                                    8.0000
3
             Sample size (N)
                                10000.0000
 Discovered Prop (out of N)
                                    0.7483
5
           Explained Variance
                                    0.5040
6
            Missing Variance -952625.2870
               Noise Variance 952625.7830
8 Theoretical Noise Std. Dev.
                                  140.3120
```

Fig. 4 Truth vs. Estimate

From Fig. 4, it is seen that the first input string (30to40MaleUKNo) is sent to the server 705 times. However, the estimated value by the decode function is 1432. Std error between these values is 255 and the corresponding proportion is 0.0255 which is well below the significance level (alpha = 0.05). It is inferred from the above results that the actual data is privacy preserved and could also effectively used for obtaining statistics.

```
$privacy
                           values
           parameters
1
          Effective p 0.5625000
          Effective q 0.6875000
2
3
             exp(e_1) 2.9279012
                   e_1 1.0742859
4
           exp(e_inf) 81.0000000 which(x, arr.ind = FA
5
                e_inf 4.3944492
6
7 Detection frequency 0.1100469
$params
                    values
  parameters
1
           k 1.600000e+01
2
           h 2.000000e+00
3
           m 8.000000e+00
4
           p 5.000000e-01
5
           q 7.500000e-01
6
           f 5.000000e-01
7
           N 1.000000e+04
8
       alpha 2.777778e-03
```

Fig. 5 Privacy parameters used

\$metrics
\$metrics\$sample_size
[1] 10000

\$metrics\$allocated_mass
[1] 0.7483

\$metrics\$num_detected
[1] 8

\$metrics\$explained_var
[1] 0.504

\$metrics\$missing_var
[1] -952625.3

Fig. 6 Results/ metrics

In Fig. 6, the metric \$num_detected represents the number of unique candidate strings obtained after decoding. When the sample size is less, this metric will mostly be zero. This means the server could not decode any survey responses and hence statistical analysis on that data cannot be performed. Therefore, it is often recommended to increase the sample size to few thousands before encoding. For the purpose of this demonstration, I have used 10000 samples. (Refer Fig. 2)

Reference:

Source code obtained from: https://github.com/google/rappor.