1.3 Pearson's Correlation Coefficient Comparison

This is to test whether the synthetic data has captured dependencies between variables of the original data.

We calculate Pearson's correlation coefficients – r, between categorical columns within the original and synthetic data.

To compare the how similar those r's are,

- We calculate the **MSE** between every pair of *r*'s (for the same pair of columns but one from synthetic one from real data)
- We calculate the **SRA** (Synthetic Ranking Agreement, see explanation below) of *r*'s for each column.

```
In [6]:
 1
   def r corr test(df,PTable = False, CoefficientandPtabl
 2
        '''Returns a table of Pearson's r correlation coef
 3
 4
       Args:
 5
       df: The input dataframe
       PTable: False (default) or True, if True, then the
 6
 7
       CoefficientandPtable: False(default) or True, if t
 8
       lower: True(default) or False. If True, the lower
 9
10
       Returns:
11
        The requested table as specified in the args. If P
12
        1.1.1
13
       from scipy.stats import pearsonr
14
15
        import pandas as pd
16
        import numpy as np
17
       df index = (df.keys()).tolist()
18
19
       n = len(df index)
20
        ini = [ [ None for y in range( n ) ]
21
                     for x in range( n ) ]
22
23
        #pearsonr returns two values: the correlation coef
24
        #so we create two empty dataframes to store them
25
        coefficient table = pd.DataFrame(ini,index = df in
26
       p table = coefficient table.copy()
27
        coe and p table = coefficient table.copy()
28
29
       for i in range(n):
30
            for j in range(i+1,n):
                name1 = df index[i]
31
                name2 = df index[j]
32
                obs_1 = df[name1].dropna()
33
                obs 2 = df[name2].dropna()
34
```

```
35
                dataframe = pd.DataFrame({name1: obs 1, na
36
                values = dataframe.dropna().values
37
                (coe,p) = pearsonr(values[:,0],values[:,1]
38
39
                coefficient table.loc[name1,name2]=coe
                p table.loc[name1,name2]=p
40
                coe and p table.loc[name1,name2]=(coe,p)
41
42
        if lower:
43
            #A function that can fill the lower part of th
44
            #But for comparison reasons you may want them
45
            def fill lower(df):
46
                n = df.values.shape[0]
47
                for j in range(n):
48
                    for i in range(j+1,n):
49
50
                        df.iloc[i,j]=df.iloc[j,i]
51
                return df
52
            coefficient table = fill lower(coefficient tab
53
            p table = fill lower(p table)
54
55
            coe_and_p_table = fill_lower(coe_and_p_table)
56
57
        if PTable:
58
59
            return p table
60
        elif CoefficientandPtable:
61
            return coe and p table
        else:
62
63
            return coefficient table
```

What is SRA?

SRA is used when we want to test whether a synthetic dataset respects a certain ranking. Suppose we have a list of metrics $R_1, R_2, R_3 \dots R_n$ calculated from the real data and a list of same metrics $S_1, S_2, S_3 \dots S_n$ calculated from the synthetic data. Then we define SRA as

$$SRA(R, S) = \frac{1}{n(n-1)} \sum_{i=1}^{n} \sum_{j \neq i} Id((S_i - S_j) \times (R_i - R_j) > 0)$$

where Id is the identity function. $SRA \in [0,1]$, The closer the SRA to 1, the better the ranking agreement.

In the case of correlation comparison, suppose we have columns A, B, C, D, E, for column A, we calculate correlation coefficients r_{AB} , r_{AC} , r_{AD} , r_{AE} for the real data, and r'_{AB} , r'_{AC} , r'_{AD} , r'_{AE} for the synthetic data. We hope the ranking of r and r' agrees, e.g. if $r_{AB} > r_{AC}$ then $r'_{AB} > r'_{AC}$ as well. As a result, our R is r_{AB} , r_{AC} , r_{AD} , r_{AE} , S is r'_{AB} , r'_{AC} , r'_{AD} , r'_{AE} , and we can calulate the SRA for each column.

```
In [7]:
   def SRA(R,S):
 1
        '''Calculate the SRA of lists R and S
 2
 3
        Args:
 5
        - R: A list of performance metrics of different pr
 6
        - S: A list of performance metrics of different pr
 7
 8
        Returns:
 9
        - SRA: SRA value
10
11
        def identity function(statement):
12
            ∨ = 0
13
            if statement:
14
15
                v = 1
16
            return v
17
18
        k = len(R)
19
        sum = 0
        for i in range(k):
20
            for j in range(k):
21
                if i != j:
22
                    if (R[i]-R[j])==0:
23
24
                         if (S[i]-S[j])==0:
25
                             agree = True
26
                         else:
27
                             agree = False
28
                     else:
                         agree = (R[i]-R[j])*(S[i]-S[j])>0
29
30
                     sum += identity function(agree)
        SRA = sum_ / (k*(k-1))
31
32
        return SRA
```

```
In [8]:
 1
   def CorrelationSRA(ori correlation df,gen correlation
        '''Returns the value of SRA for the absolute Pears
 2
 3
        all other columns. SRA is between 0 and 1, the clo
        the more similar the synthetic data and the real d
 4
 5
 6
       Args:
        ori correlation df: the correlation coefficient da
 7
 8
                             r corr test.
        gen correlation df: the correlation coefficient da
 9
10
                             r corr test.
       ColumnWise: False(default) or True. If True, the r
11
12
                    Otherwise, the return is the average o
13
14
        Returns:
15
        s: It is either a column-wise SRA series or the av
16
        I = I = I
17
        import numpy as np
18
19
        import pandas as pd
20
       columns = (ori correlation df.keys()).tolist()
21
       n = len(columns)
22
       ini = np.ones(n)
23
24
25
       for i in range(n):
            ori values = ori correlation df.iloc[i,:].drop
26
27
            gen values = gen correlation df.iloc[i,:].drop
28
            ini[i] = SRA(abs(ori values), abs(gen values))
29
        if ColumnWise:
30
            s = pd.Series(ini,index = columns)
31
            s['average'] = sum(ini)/n
32
        else:
33
34
            s = sum(ini)/n
```

```
35
       return s
   def MSE(r table ori,r table gen):
 1
 2
 3
       Returns the MSE for each position between two data
 5
       import pandas as pd
 6
       import numpy as np
       ori = r table ori.fillna(0).values
 7
       gen = r table gen.fillna(0).values
 8
       columns = (r_table_gen.keys()).tolist()
9
       matrix = (ori-gen)**2
10
       df = pd.DataFrame(matrix, index = columns, columns
11
       score = np.sum(matrix)/(len(ori)*(len(ori)-1)) #Th
12
       return df, score
13
```

Data Loading

```
In [32]:
   import numpy as np
 1
 2
   import pandas as pd
   dp ori df = pd.read csv('synthetic data/doppelGANger/d
 3
   dp gen df = pd.read csv('synthetic data/doppelGANger/d
 4
   tgan ori df = pd.read csv('synthetic data/TGAN/tgan or
 5
   tgan gen df = pd.read csv('synthetic data/TGAN/tgan ge
 6
   ori df = pd.read csv('synthetic data/2 no id/ori df.cs
 7
   gen 1 df = pd.read csv('synthetic data/2 no id/gen 1 d
 8
   gen 2 df = pd.read csv('synthetic data/2 no id/gen 2 d
 9
   gen 3 df = pd.read csv('synthetic data/2 no id/gen 3 d
10
   gen 4 df = pd.read csv('synthetic data/2 no id/gen 4 d
11
12
   synthetic data dic = {'DoppelGANger ini':[dp ori df, d
13
                          'gen 1':[tgan ori df,gen 1 df],'
14
                         'gen 4':[tgan ori df,gen 4 df]}
15
   syn keys = list(synthetic data dic.keys())
16
```

```
In [37]:
 1 \mid n = len(syn keys)
   MSE array = np.zeros(n)
 2
   for i in range(n):
 3
        key = syn keys[i]
 4
 5
        df ori = synthetic data dic[key][0]
        df gen = synthetic data dic[key][1]
 6
 7
        r table ori = r corr test(df ori)
        r table gen = r corr test(df gen)
 8
 9
10
        #Highlight all r values > 0.5 as yellow, indicating
        def color threshold yellow(val):
11
            threshold = 0.5
12
            if ((val != None) and (abs(val) > threshold)):
13
                color = 'yellow'
14
15
            else:
                color = 'black'
16
            return 'color: %s' % color
17
18
19
        display(key+' '+'generated r table', r table gen.st
        display(key+' '+'real r table',r table ori.style.a
20
        sra = CorrelationSRA(r table ori,r table gen,Colum
21
        if i==0:
22
            sra df = pd.DataFrame(sra,columns = [key])
23
24
        else:
25
            sra df = pd.concat([sra df,pd.DataFrame(sra,co]
        display(kev+' '+'SRA',sra)
26
        MSE df, MSE score = MSE(r table gen,r table ori)
27
        display(kev+' '+'MSE table', MSE df)
28
        MSE array[i] = MSE score
29
30
   MSE series = pd.Series(MSE array,index = syn keys)
 'DoppelGANger ini generated r table'
```

	dday	weight	height	age	temp
dday	None	0.460590	0.579901	0.385757	0.572924
weight	0.460590	None	0.852360	0.765801	0.725225
height	0.579901	0.852360	None	0.700063	0.951644
age	0.385757	0.765801	0.700063	None	0.561933
temp	0.572924	0.725225	0.951644	0.561933	None
'Dopp	elGANger	ini real	r table'		
	dday	weight	height	age	temp
dday	None	0.547442	0.625742	0.431480	0.624604
weight	0.547442	None	0.904009	0.888397	0.787127
height	0.625742	0.904009	None	0.739106	0.964485
age	0.431480	0.888397	0.739106	None	0.589636
temp	0.624604	0.787127	0.964485	0.589636	None
dday weigh heigh age temp avera	t 1.0 1.0 1.0 ge 1.0 e: float64))))))	table'		
	dday	weight	height	age	temp
dday	0.000000	0.007543	0.002101	0.002091	0.002671
weight	0.007543	0.000000	0.002668	0.015030	0.003832
height	0.002101	0.002668	0.000000	0.001524	0.000165
	I				

0.002091 0.015030 0.001524 0.000000 0.000767

0.002671 0.003832 0.000165 0.000767 0.000000

age

temp

	dday	height	weight	temp	vomit_dur
dday	None	0.630282	0.674201	-0.518343	0.509896
height	0.630282	None	0.967306	0.271986	0.840133
weight	0.674201	0.967306	None	0.264717	0.934518
temp	-0.518343	0.271986	0.264717	None	0.417741
vomit_dur	0.509896	0.840133	0.934518	0.417741	None
cough_dur	0.106260	0.607276	0.704238	0.737285	0.848396
diar_No	-0.508896	-0.476903	-0.642798	-0.165282	-0.777692
diar_Yes	0.505435	0.444277	0.610614	0.136410	0.744161
head_No	0.013767	-0.627081	-0.679343	-0.845591	-0.802912
head_Yes	-0.013388	0.627144	0.679569	0.845392	0.803213

^{&#}x27;tGAN real r table'

		dday	height	weight	temp	vomit_dur
dday		None	0.176702	0.216054	-0.090951	-0.039075
height		0.176702	None	0.873474	-0.143156	-0.026711
weight		0.216054	0.873474	None	-0.122154	-0.018737
temp		-0.090951	-0.143156	-0.122154	None	0.125559
vomit_du	r	-0.039075	-0.026711	-0.018737	0.125559	None
cough_d	ır	-0.091015	-0.059380	-0.041915	0.112293	0.020258
diar_No		0.017397	0.053020	0.015691	-0.048428	-0.147209
diar_Yes		-0.017397	-0.053020	-0.015691	0.048428	0.147209
head_No		0.069809	-0.108343	-0.093078	-0.281744	-0.086867
head_Ye	6	-0.069809	0.108343	0.093078	0.281744	0.086867

^{&#}x27;tGAN SRA'

```
dday
          0.694444
height
          0.638889
weight
          0.555556
temp
          0.750000
vomit_dur
          0.166667
cough_dur 0.583333
diar_No
          0.500000
         0.500000
diar_Yes
head_No
          0.805556
head_Yes 0.805556
average 0.600000
dtype: float64
'tGAN MSE table'
```

	dday	height	weight	temp	vomit_dur	col
dday	0.000000	0.205735	0.209899	0.182664	0.301369	0.0
height	0.205735	0.000000	0.008805	0.172343	0.751419	0.44
weight	0.209899	0.008805	0.000000	0.149669	0.908696	0.5
temp	0.182664	0.172343	0.149669	0.000000	0.085370	0.39
vomit_dur	0.301369	0.751419	0.908696	0.085370	0.000000	0.68
cough_du	r 0.038918	0.444430	0.556745	0.390614	0.685812	0.00
diar_No	0.276985	0.280819	0.433608	0.013655	0.397508	0.50
diar_Yes	0.273354	0.247304	0.392258	0.007741	0.356351	0.4
head_No	0.003141	0.269089	0.343707	0.317924	0.512721	0.70
head_Yes	0.003183	0.269154	0.343971	0.317700	0.513152	0.70

'gen 1 generated r table'

	dday	height	weight	temp	vomit_dı
dday	None	0.162797	0.097409	-0.339070	0.110246
height	0.162797	None	0.937405	-0.563485	-0.075009
weight	0.097409	0.937405	None	-0.514464	-0.047376
temp	-0.339070	-0.563485	-0.514464	None	0.589742
vomit_du	0.110246	-0.075009	-0.047376	0.589742	None

	dday	height	weight	temp	vomit_dı
cough_dur	0.021433	-0.798786	-0.693646	0.729940	0.560289
diar_No	0.225388	0.606384	0.488053	-0.108582	0.235987
diar_Yes	-0.242572	-0.623032	-0.500043	0.125227	-0.25027
head_No	0.366513	-0.024406	-0.007414	-0.748058	-0.64679
head_Yes	-0.368334	0.024948	0.008275	0.745608	0.647633
'gen 1	real r tab	le'			
	dday	height	weight	temp	vomit_dur
dday	None	0.176702	0.216054	-0.090951	-0.039075
height	0.176702	None	0.873474	-0.143156	-0.026711
weight	0.216054	0.873474	None	-0.122154	-0.018737
temp	-0.090951	-0.143156	-0.122154	None	0.125559
vomit_dur	-0.039075	-0.026711	-0.018737	0.125559	None
cough_dur	-0.091015	-0.059380	-0.041915	0.112293	0.020258
diar_No	0.017397	0.053020	0.015691	-0.048428	-0.147209
diar_Yes	-0.017397	-0.053020	-0.015691	0.048428	0.147209
head_No	0.069809	-0.108343	-0.093078	-0.281744	-0.086867
head_Yes	-0.069809	0.108343	0.093078	0.281744	0.086867
'gen 1	SRA'				
dday height	0.33 0.52	3333			
weight	0.52				
temp	0.86				
	dur 0.61				
cough_c					
diar_No diar_Ye	0.52 es 0.55				
head_No					
head_Ye					
average	0.58	8889			
dtype:	float64				
I					

	dday	height	weight	temp	vomit_dur	col
dday	0.000000	0.000193	0.014077	0.061563	0.022297	0.0
height	0.000193	0.000000	0.004087	0.176676	0.002333	0.54
weight	0.014077	0.004087	0.000000	0.153907	0.000820	0.4;
temp	0.061563	0.176676	0.153907	0.000000	0.215465	0.38
vomit_dur	0.022297	0.002333	0.000820	0.215465	0.000000	0.29
cough_dur	0.012645	0.546722	0.424753	0.381487	0.291634	0.00
diar_No	0.043260	0.306211	0.223126	0.003619	0.146840	0.18
diar_Yes	0.050704	0.324913	0.234597	0.005898	0.157996	0.18
head_No	0.088033	0.007046	0.007338	0.217449	0.313523	0.0
head_Yes	0.089117	0.006955	0.007192	0.215170	0.314458	0.0

'gen 2 generated r table'

	dday	height	weight	temp	vomit_dur
dday	None	0.669932	0.600616	-0.421997	-0.319321
height	0.669932	None	0.936421	-0.027013	0.160809
weight	0.600616	0.936421	None	-0.117218	0.049449
temp	-0.421997	-0.027013	-0.117218	None	0.861116
vomit_dur	-0.319321	0.160809	0.049449	0.861116	None
cough_dur	-0.365681	-0.505931	-0.566673	0.197196	0.291194
diar_No	0.303687	-0.073868	0.027920	-0.615479	-0.684354
diar_Yes	-0.324059	0.053488	-0.042256	0.611958	0.660524
head_No	0.355109	-0.323211	-0.260020	-0.551272	-0.723152
head_Yes	-0.355177	0.323054	0.259797	0.551744	0.723586

^{&#}x27;gen 2 real r table'

1					
	dday	height	weight	temp	vomit_dur
dday	None	0.176702	0.216054	-0.090951	-0.039075
height	0.176702	None	0.873474	-0.143156	-0.026711
weight	0.216054	0.873474	None	-0.122154	-0.018737
temp	-0.090951	-0.143156	-0.122154	None	0.125559
vomit_dur	-0.039075	-0.026711	-0.018737	0.125559	None
cough_dur	-0.091015	-0.059380	-0.041915	0.112293	0.020258
diar_No	0.017397	0.053020	0.015691	-0.048428	-0.147209
diar_Yes	-0.017397	-0.053020	-0.015691	0.048428	0.147209
head_No	0.069809	-0.108343	-0.093078	-0.281744	-0.086867
head_Yes	-0.069809	0.108343	0.093078	0.281744	0.086867
'gen 2 dday height weight	0.69	1111 4444 5556			
temp		1111			
cough_d diar_No diar_Ye head_No head_Ye average dtype:	o.50 0.83 0.80	3333 5556 7778 7778			
	dday	height	weight	temp von	nit_dur cı
	uuay		147000 0.1	-	nit_dur c

	dday	height	weight	temp	vomit_dur	CI
dday	0.000000	0.243275	0.147888	0.109591	0.078538	0.
height	0.243275	0.000000	0.003962	0.013489	0.035164	0.
weight	0.147888	0.003962	0.000000	0.000024	0.004649	0.
temp	0.109591	0.013489	0.000024	0.000000	0.541044	0.
vomit_dur	0.078538	0.035164	0.004649	0.541044	0.000000	0.

	dday	height	weight	temp	vomit_	_dur	CI
cough_dur	0.075441	0.199408	0.275371	0.007208	0.0734	-06	0.
diar_No	0.081962	0.016101	0.000150	0.321547	0.2885	525	0.
diar_Yes	0.094041	0.011344	0.000706	0.317566	0.2634	92	0.
head_No	0.081396	0.046168	0.027870	0.072646	0.4048	359	0.
head_Yes	0.081435	0.046101	0.027795	0.072900	0.4054	10	0.
'gen 3 generated r table'							
	dday	height	weight	temp	V	omit_c	dur
dday	None	-0.162625	-0.23094	9 0.0903	84 0.	04743	34
height	-0.162625	None	0.767620	0.2500	86 0.	004119	9
weight	-0.230949	0.767620	None	0.24196	61 –0	0.10198	85
temp	0.090384	0.250086	0.241961	None	0.	07435	8
vomit_dur	0.047434	0.004119	-0.101985	0.0743	58 No	one	
cough_dur	-0.064280	-0.098146	-0.03690	8 0.11335	51 –(0.0300)81
diar_No	0.083909	0.314467	0.213542	0.35189	52 –0	0.0511	09
diar_Yes	-0.083909	-0.314467	-0.213542	2 –0.3518	352 0.	051109	9
head_No	0.072363	0.170356	0.138130	0.2934	00 –0	0.0473	358
head_Yes	-0.072363	-0.170356	-0.138130	0.293	400 0.	04735	8
'gen 3	real r tak	ole'					
	dday	height	weight	temp	V	omit_	dι
dday	None	-0.036634	0.000878	-0.059	368 –	0.029	94 ⁻
height	-0.036634	None	0.881265	-0.175	099 –	0.059	27
weight	0.000878	0.881265	None	-0.160	768 –	0.040	65
temp	-0.059368	-0.175099	-0.16076	8 None	0	.17832	28
vomit_dur	-0.029941	-0.059278	-0.04065	52 0.1783	28 N	one	
cough_dur	-0.051150	-0.101346	-0.06372	9 0.1336	77 0	.04572	22

	dday	height	weight	temp	vomit_dı
diar_No	0.021721	0.105971	0.055076	-0.051018	-0.172186
diar_Yes	-0.021721	-0.105971	-0.055076	0.051018	0.172186
head_No	0.048467	-0.149711	-0.139772	-0.200484	-0.056125
head_Yes	-0.048467	0.149711	0.139772	0.200484	0.056125
'gen 3	SRA'				
dday	0.36	1111			
height	0.75	0000			
weight	0.63	8889			
temp	0.44	4444			
vomit_c	dur 0.55	5556			
cough_c	dur 0.83	3333			
diar_No	0.58	3333			
diar_Ye	es 0.61	1111			
head_No	0.63	8889			
head_Ye	es 0.61	1111			
average	0.60	2778			
dtype:	float64				
	MCE + 13 1				
'gen 3	MSE table'				

	dday	height	weight	temp	vomit_dur	col
dday	0.000000	0.015874	0.053744	0.022426	0.005987	0.00
height	0.015874	0.000000	0.012915	0.180783	0.004019	0.00
weight	0.053744	0.012915	0.000000	0.162191	0.003762	0.00
temp	0.022426	0.180783	0.162191	0.000000	0.010810	0.00
vomit_dur	0.005987	0.004019	0.003762	0.010810	0.000000	0.00
cough_dur	0.000172	0.000010	0.000719	0.000413	0.005746	0.00
diar_No	0.003867	0.043471	0.025112	0.162304	0.014660	0.00
diar_Yes	0.003867	0.043471	0.025112	0.162304	0.014660	0.00
head_No	0.000571	0.102443	0.077229	0.243921	0.000077	0.0
head_Yes	0.000571	0.102443	0.077229	0.243921	0.000077	0.0
'gen 4	generated	r table'				

	dday	height	weight	temp	vomit_dur
dday	None	0.725814	0.780287	-0.703732	-0.398647
height	0.725814	None	0.903226	-0.237532	-0.087795
weight	0.780287	0.903226	None	-0.289294	-0.045537
temp	-0.703732	-0.237532	-0.289294	None	0.486734
vomit_dur	-0.398647	-0.087795	-0.045537	0.486734	None
cough_dur	-0.284812	-0.554406	-0.315996	0.058875	0.440218
diar_No	0.345047	0.303831	0.183392	-0.363811	-0.142562
diar_Yes	-0.367659	-0.318263	-0.198528	0.400465	0.150752
head_No	0.300212	-0.093038	-0.116846	-0.476977	-0.921556
head_Yes	-0.299526	0.093150	0.117114	0.476242	0.921134
'gen 4	real r tab		woisht	tomp	vomit du
	dday	height	weight	temp	vomit_dur
dday	None	0.176702	0.216054	-0.090951	-0.039075
height	0.176702	None	0.873474	-0.143156	-0.026711
weight	0.216054	0.873474	None	-0.122154	-0.018737
temp	-0.090951	-0.143156	-0.122154	None	0.125559
vomit_dur	-0.039075	-0.026711	-0.018737	0.125559	None
cough_dur	-0.091015	-0.059380	-0.041915	0.112293	0.020258
diar_No	0.017397	0.053020	0.015691	-0.048428	-0.147209
diar_Yes	-0.017397	-0.053020	-0.015691	0.048428	0.147209
head_No	0.069809	-0.108343	-0.093078	-0.281744	-0.086867
head_Yes	-0.069809	0.108343	0.093078	0.281744	0.086867
'gen 4	SRA'				
'gen 4 dday		1111			
		4444			

temp		0.527778
vomit	_dur	0.555556
cough	_dur	0.444444
diar_	No	0.583333
diar_	Yes	0.583333
head_	No	0.694444
head_	Yes	0.694444
avera	ge	0.608333
dtype	: float6	54
'gen	4 MSE ta	ıble'

5	l .	•	 _	_	-	_	_	

	dday	height	weight	temp	vomit_dur	col
dday	0.000000	0.301524	0.318359	0.375500	0.129292	0.0
height	0.301524	0.000000	0.000885	0.008907	0.003731	0.24
weight	0.318359	0.000885	0.000000	0.027936	0.000718	0.0
temp	0.375500	0.008907	0.027936	0.000000	0.130447	0.00
vomit_dur	0.129292	0.003731	0.000718	0.130447	0.000000	0.17
cough_dur	0.037557	0.245051	0.075120	0.002854	0.176366	0.00
diar_No	0.107354	0.062906	0.028124	0.099466	0.000022	0.0
diar_Yes	0.122683	0.070354	0.033429	0.123930	0.000013	0.0
head_No	0.053085	0.000234	0.000565	0.038116	0.696706	0.0
head_Yes	0.052770	0.000231	0.000578	0.037830	0.696002	0.0

```
In [34]:
 1 display('MSE values of r for each synthetic data', MSE_
 'MSE values of r for each synthetic data'
 DoppelGANger ini
                   0.003839
 tGAN
                     0.315190
 gen 1
                     0.129868
                     0.111499
 gen 2
                     0.048782
 gen 3
                     0.098235
 gen 4
 dtype: float64
```

Conclusion

- The lower the value of MSE, the smaller the average difference between the correlation coefficients between the real and synthetic data, hence the better the result.
- DoppelGANger ini is good but it contains only 5 columns.
- tGAN, gen 1 and gen 2 have large MSE values, by comparing the correlation tables we find that these generated datas have wronly too strong dependencies (r>0.5) between some columns.
- gen 3 agrees with its original data in all r>0.5.

In [35]:

display('SRA for each column and synthetic data',sra_d

^{&#}x27;SRA for each column and synthetic data'

	DoppelGANger ini	tGAN	gen 1	gen 2	gen 3
dday	1.0	0.694444	0.333333	0.861111	0.361111
weight	1.0	0.555556	0.527778	0.805556	0.638889
height	1.0	0.638889	0.527778	0.694444	0.750000
age	1.0	NaN	NaN	NaN	NaN
temp	1.0	0.750000	0.861111	0.361111	0.444444
average	1.0	0.600000	0.588889	0.666667	0.602778
vomit_du	ır NaN	0.166667	0.611111	0.750000	0.55556
cough_d	ur NaN	0.583333	0.500000	0.500000	0.833333
diar_No	NaN	0.500000	0.527778	0.833333	0.583333
diar_Yes	NaN	0.500000	0.555556	0.805556	0.611111
head_No	NaN	0.805556	0.722222	0.527778	0.638889
head_Yes	s NaN	0.805556	0.722222	0.527778	0.611111

Conclusion

- we can conclude that DoppleGANger ini preserves the dependency ranking between columns very well.
- By comparing the 'average', gen 2 is best at preserve the ranking. Though in the previous section we find that gen 2 tends to have a large r value in average.

Possible Improvements in this method

Note in the TGAN data, it contains categorical columns e.g. diar_No and diar_Yes between which the r is -1. This corresponding to the logic fact that if diar_No = 1, then diar_Yes = 0; if diar_No = 0, then diar_Yes = 1. A reasonable synthetic data has to respect this kind of 'logic' relationship.

As a result, it's reasonable to say that that larger the absolute value of r is, the more important the relationship is, that's why we choose to use MSE rather than MAE in quantitative evaluation, a possible improvement is adding weight according to r rather than averaging the MSE.

Reference:

 James Jordon, Jinsung Yoon, Mihaela van der Schaar. PATE-GAN: GENERATING SYNTHETIC DATA WITH DIFFERENTIAL PRIVACY GUARANTEES (https://openreview.net/pdf?id=S1zk9iRqF7)