



Big Data Systems

# REDIS - MONGODB

## ASSIGNMENT

## TASK 1 | RESULTS

- 1.1. How many users modified their listing on January?

```
> r$BITCOUNT("ModificationsJanuary")  
[1] 9969
```

- 1.2. How many users did NOT modify their listing on January?

Note that the sum of the numbers occurred from task 1.1 and 1.2 (20,000) do not match the total number of users (19,999). This happens because all BITOP operations occur at byte-level increments. Each byte has 8 bits and the results of a BITOP operation should always be an integer multiple of 8.

```
> r$BITCOUNT("results1")  
[1] 10031
```

- 1.3. How many users received at least one e-mail per month (at least one e-mail in January and at least one e-mail in February and at least one e-mail in March)?

```
> r$BITCOUNT("results2")  
[1] 16006
```

- 1.4. How many users received an e-mail on January and March but NOT on February?

```
> r$BITCOUNT("combined result")  
[1] 2417
```

- 1.5. How many users received an e-mail on January that they did not open but they updated their listing anyway?

```
> r$BITCOUNT("EmailsopenedJanuary")  
[1] 2807  
> |
```

- 1.6. How many users received an e-mail on January that they did not open but they updated their listing anyway on January OR they received an e-mail on February that they did not open but they updated their listing

anyway on February OR they received an e-mail on March that they did not open but they updated their listing anyway on March?

```
[1] 7221  
> |
```

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- 1.7. Does it make any sense to keep sending e-mails with recommendations to sellers? Does this strategy really work? How would you describe this in terms a business person would understand?

After analysing the data, it seems that the specific strategy of the classified provider to send personalized e-mails has a partial effect on motivating the sellers to improve their listings. This conclusion can be derived from the fact that almost 50% of these sellers during this 3-month period proceeded to a change of their listings after having received and opened the respective recommendation e-mail (January: 49.5%, February: 50.2%, March: 49.9%). Therefore, in case the provider is willing to use the specific approach as a dominant means to have earnings, what should be improved to draw the attention of the recipients of these e-mails more efficiently should be investigated.

## TASK 2 | RESULTS

For this task before adding the data files in MongoDB, a function is created in R via which the below manipulation/cleaning procedure is followed for several fields:

- Motorcycles price (`ad_data$Price`): Initially defined as string. For the cases where:
  - A number existed in this field, it is converted to a numeric data type and the currency symbol is removed.
  - The price is relatively small (i.e. 10 euros), we consider that these do not probably consist cases of motorcycle sale (for example they could be motorcycle parts).
  - For the cases where no price is defined (i.e. `AskForPrice`), we introduce a counter to measure the number of these motorcycles.
- Registration date (`ad_data$Registration`): Remove the month from the date field
- Mileage (`ad_data$Mileage`): Remove the measurement unit of the motorcycles' mileage (i.e. km)
- Advertisement age (`ad_data$Ad_age`): Introduced this new field to count the age of the advertisement

2.2 How many bikes are there for sale?

```
[1] 29701
> |
```

2.3 What is the average price of a motorcycle (give a number)? What is the number of listings that were used in order to calculate this average (give a number as well)? Is the number of listings used the same as the answer in 1.2? Why?

The motorcycles which are not taken into account to calculate this result are these for which no price is defined (i.e. AskForPrice = TRUE or Price=NA).

```
_id AvgPrice count
1 NA 3007.571 28710
> |
```

2.4 What is the maximum and minimum price of a motorcycle currently available in the market?

```
_id MinPrice MaxPrice
1 NA 10 89000
> |
```

2.5 How many listings have a price that is identified as negotiable?

```
_id count
1 NA 1348
> |
```

2.6 For each Brand, what percentage of its listings is listed as negotiable?  
(indicative results)

```
+ )
      _id TotalCount NegotiableCount NegotiableRatio
1      Adler          1             0      0.000000
2      Jinlun          1             1     100.000000
3  Bombardier          1             1     100.000000
4      Genata          2             0      0.000000
5      Barossa          1             0      0.000000
6      Qingqi          2             2     100.000000
7      Fever          1             1     100.000000
8      Ariel           2             0      0.000000
9      Aie             4             0      0.000000
10     Vedim           1             0      0.000000
11     Rewaco          3             1     33.333333
12  Super Moto          2             1     50.000000
13     Eagle           3             0      0.000000
14     Adiva           5             4     80.000000
15     Niu             1             1     100.000000
16     Maico           4             1     25.000000
17     Goes            1             0      0.000000
18     Bultaco          2             1     50.000000
19     Dias            3             3     100.000000
20     Harlow           1             0      0.000000
21     Amstrong          3             3     100.000000
22     Zuendapp          7             3     42.857143
```

2.7 What is the motorcycle brand with the highest average price?

```
      _id AvgPrice count
1 semog      15600      1
> |
```

2.8 What is the TOP 10 models with the highest average age? (Round age by one decimal number)

```
      _id AvgAge count
1 Αλλο henderson indian replica      88.0      1
2              Bmw R 12                85.0      1
3 Αλλο MATCHLESS G3 350                84.0      1
4              Norton H16               83.0      1
5 Αλλο Matsoules                       81.0      1
6 Αλλο Matchless G3/L                  80.0      1
7 Αλλο NEW HUDSON                      80.0      1
8 Bsa M20 ARMY MOTO!                   80.0      1
9              Bsa M20                   79.5      2
10             Norton 16H                79.0      1
```

2.9 How many bikes have "ABS" as an extra?

```
_id count
1 NA 4025
> |
```

2.10 What is the average Mileage of bikes that have "ABS" AND "Led lights" as an extra?

```
_id AvgMilage count
1 NA 30125.7 1135
> |
```

2.11 Which is the TOP 3 colors per bike category? (indicative results)

_id	colors
1	Goes orange (Metallic), 1
2	Niu Black, 1
3	Shandong Liangzi other, 1
4	Adler Red, 1
5	AMS Blue, Green, Yellow, 1, 1, 1
6	New Force Motor Black, 1
7	Wsk Orange, 1
8	Eagle Red, Orange (Metallic), white (Metallic), 1, 1, 1
9	Nitro Motors Dark blue, 1
10	Beeline Black, 1
11	KL Black, 1
12	Bombardier Yellow, 1
13	Laverda Silver, orange, 1, 1
14	Harlow Black (Metallic), 1
15	Morini Light Blue (Metallic), Blue (Metallic), Red (Metallic), 1, 1, 1
16	Genata Red, Silver, 1, 1
17	Bultaco Dark blue, Light Blue, 1, 1
18	Vedim Black, 1
19	E-Ton Red, Black, 1, 1
20	Boatian Black, Silver, Blue, 1, 1, 1
21	Kinroad Green, Yellow, Grey, 1, 1, 1
22	Ural white, Beige, Blue (Metallic), 1, 1, 1
23	Hartford Dark Red, 1
24	Generic Yellow, orange, Black (Metallic), 1, 1, 1
25	Chang Jiang Silver, 1
26	Odess white, Black, 1, 1
27	Amstrong Yellow, Red, Black, 1, 1, 1
28	Dias Blue (Metallic), white, Red, 1, 1, 1
29	Fever Orange, 1
30	Victory Bordeaux (Metallic), 1
31	Buggy Motors Black, Black (Metallic), Red, 1, 1, 1
32	Barossa Light Blue, 1