# 1 - SQL tasks

1. Report what is the average salary of men and women in the company, for all time?

Answer:

For women : **61432.56347245** For men : **61464.61142705** 

2. What year and month did the company hire the most people?

Answer: 03/1985 (number of people hired: 3431)

3. What is the expected percentage increase in salary that an employee can expect after working in the company for 3 years?

Answer: 8.095 %

- 4. Which employee had the biggest increase in salary during his time in the company? Answer: ( I understand it in two different ways, so I answered both cases )
  - The employee that had the biggest increase in salary in term of percentage in comparison with his starting salary is the employee **484934**, his salary increased by : **129.40** % in comparison with his starting salary.
  - The employee that had the biggest increase in salary in term of how much did his salary increase is the employee **43145**, his salary increased by : **53875** in comparison with his starting salary.
- 5. Which month had the biggest difference between the lowest and highest salary in the company?

Answer: 2002/05

6. Let's say that every salesperson sells 500 products per month. Each sale is making the company 300 EUR in profit. What is the profit or loss of the company for each month. Assume that the only profit are products sold by salespeople and the only cost the salaries of all employees.

Answer:

result year	month	1
-417687	1985	1
-44728884	1985	2
-11135736	1985	3
33673616	1985	4
71358847	1985	5
114504625	1985	6
152531887	1985	7
194406737	1985	8
237667187	1985	9

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283886045
            1985
                  10
328534457
            1985
                  11
372704031
            1985
                  12
413710174
            1986
                  1
376254207
            1986
                  2
398345094
            1986
447743232
            1986
                  4
477701903
            1986
                  5
528519594
            1986
                  6
558500362
            1986
                  7
600113630
            1986
                  8
647584853
            1986
                  9
685994058
            1986
                  10
726646717
            1986
                  11
764840235
            1986
                  12
816803463
            1987
                  1
780442871
            1987
                  2
791218802
            1987
                  3
844730642
            1987
                  4
871268670
            1987
                  5
925149200
            1987
                  6
946022360
            1987
                  7
988348933
            1987
                  8
1037770182
            1987
                  9
1066419305
            1987
                  10
1107995216 1987
                  11
1144349302 1987
                  12
1198964574 1988
                  1
1159156301
            1988
                  2
1176274375 1988
                  3
1226352186 1988
                  4
1252663805
            1988
                  5
1310605159 1988
                  6
1324345204 1988
                  7
1369576467
            1988
                  8
1428075847 1988
                  9
1450525543
            1988
                  10
1491150559
            1988
                  11
1519311292 1988
                  12
1573192702 1989
                  1
1548033628 1989
                  2
1547253607
                  3
            1989
1595681495 1989
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1626691170 1989 5
1690278046 1989
                 6
1700419870 1989
                 7
1742255865 1989
                 8
1804224232 1989
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1822639417 1989
                 10
1866965362 1989
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1893204167 1989
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1952023989 1990
                 1
1926554556 1990
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1915184944 1990
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1969755453 1990
                 4
1993357426 1990
2060883148 1990
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2063189515 1990
                 7
2095917915 1990
2167013264 1990
                 9
2180180816 1990
                 10
2228898334 1990
                 11
2256094249 1990
                 12
2312007025 1991
                 1
2294186144 1991
                 2
2274509183 1991
                 3
2330169040 1991
                 4
2339231176 1991
2412896135 1991
2416799966 1991
                 7
2439606571 1991
                 8
2515316761 1991
                 9
2526860702 1991
                 10
2568967571 1991
                 11
2591295314 1991
                 12
2651397646 1992
                 1
2619492795 1992
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2608695853 1992
2672346675 1992
2680991146 1992
2757398246 1992
2749246649 1992
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2774315186 1992 8
2862277594 1992
2860496337
           1992
                 10
2907039287 1992
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2926979976 1992
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2972973066 1993
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2974832971 1993
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2933677177 1993
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2997622402 1993
3001648282 1993
3076767483 1993
3072809061 1993
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3096986658 1993
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3180154918 1993
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3178053081 1993
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3215409625 1993
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3227656222 1993
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3278480484 1994
                 1
3287426271 1994
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3237883803 1994
3305976827 1994
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3299813111 1994
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3384124894 1994
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3381484660 1994
3408775791 1994
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3498823600 1994
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3493710749 1994
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3538224515 1994
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3553075182 1994
                 12
3606473140 1995
                 1
3620066233 1995
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3560790569 1995
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3630288555 1995
3624021817 1995
3700267968 1995
                 6
3691347802 1995
3706742539 1995 8
3807359858 1995
                 9
3787318594 1995
                 10
3832343786 1995
                 11
3838411675 1995
                 12
3887525126 1996
                 1
3872621038 1996
                 2
3836575059 1996
                 3
3906523706 1996
3891822968 1996
                 5
3966385379 1996
                6
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3944422046 1996 7
3963597284 1996
                 8
4067969815 1996
                 9
4047115421 1996
                 10
4088305399 1996
                 11
4076449022 1996
                 12
4125247280 1997
                 1
4165014690 1997
                 2
4073591605 1997
                 3
4146674308 1997
4123527670 1997
4203134481 1997
4175645844 1997
4200634121 1997
                 8
4303162465 1997
                 9
4281664096 1997
4327892390 1997
                 11
4318749269 1997
                 12
4366112183 1998
4410022342 1998
                 2
4317270752 1998
                 3
4400540403 1998
                 4
4369969845 1998
4446675671 1998
                 6
4412562011
           1998
                 7
4429112953 1998
                 8
4536076424 1998
                 9
4506740221 1998
                 10
4549456776 1998
                 11
4533864948 1998
                 12
4585425192 1999
                 1
4632529048 1999
                 2
4530543955 1999
                 3
4604600739 1999
                 4
4576823422 1999
4659210798 1999
                 6
4614981247
           1999
                 7
4638358381
           1999
4749943288 1999
                 9
4717350651 1999
                 10
4765434428 1999
                 11
4746689989 1999
                 12
4793105364 2000
                 1
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4832677190 2000 2
4735195595 2000 3
4766313590 2000
                 4
4689292648 2000
4731697168 2000
4651688466 2000
                 7
4615244192 2000
4688096926 2000
                 9
4610899566 2000
                10
4611629411 2000
                 11
4545242281 2000
                 12
4549747150 2001
                 1
4639590466 2001
                 2
4496005254 2001
                 3
4533988539 2001
                 4
4454168893 2001
4494706927 2001
4408776820 2001
                 7
4378061091 2001
4454107242 2001
                 9
4378699510 2001
                 10
4378701487 2001
                 11
4316659313 2001
                 12
4318640951 2002
                 1
4399008526 2002 2
4260039913 2002 3
4298980806 2002 4
4213599221 2002 5
4259083316 2002 6
4179821131 2002 7
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5606578520 2002 8

## 2 - Data science task

Here you can use any tool of your choosing.

Please provide us with the code and the explanation of your reasoning.

1. Let's say a Senior Engineer is looking to join the company in August 2002 He wants to know how much he will be paid in 10 years assuming the company grows at the same rate as before. How would you solve this? Build a model that predicts his salary in 10 years.

### **Solution:**

The goal here is to predict the salary in 10 years of a senior engineer that will join the company at August 2002 .

According to what are we going to predict this salary?

Well, the response to this question is : depending on what informations we have about this employee. In the question we have just the date he will join the company and after how many years he wants to predict his salary.

So here, I see two solutions:

- 1) A solution that will predict his entry salary and his salary after 10 years.
- 2) A solution that will predict how much his initial salary (that should be given) will increase.

I choose to work on the first one since it includes the evolution of the salary in terms of experience.

To build a model for this problem, we should start by **building a table with the data we will need from our SQL tables**, **analyse the data we have**, **train our model** and then **generate the predictions**.

#### 1 - Create a table from the sql tables we have on the database :

The first thing that comes to my mind is to create a table that contains only employees who joined the company as senior engineers.

This tables contains the following features:

**Emp\_no**: Employee number **Gender**: Employee gender

**Hire\_date**: The date we hired the employee

**Experience:** how many years the employee has been in the company

**Salary**: the employee's salary.

So, our table: **SE\_employees**, contains the salaries of the employees for each year. (In the file BD.sql you'll find how I created this table. BD.sql should be executed to construct this table).

I also created another table called **SE\_employees\_entry** that contains the salaries of the employees at their entry. I created it in order to see if there is a correlation between the year of recruitment and the salary.

### 2 - Data Analysis:

In this step, I tried to draw some graphics to see if I can find any correlation between the features and the salary.

At my notebook, you can see that I tried to plot the salary in terms of the hiring year. I noticed that the salary doesn't increase or decrease in terms of the year. So I conclude that maybe there are other features that can determine the entry salary of a senior engineer.

Also, I plotted the salaries of each employee based on their experience in the company. I noticed that the salary increases most of the time. So, the "**experience**" feature ( which represents how many years the employee is in the company ) can be a good feature in our model.

I also considered the "gender" as a feature that I will give to my model.

#### 3 - Model training:

Before choosing a model for my solution and training it, I did the one hot encoding for the "gender" column, so I can convert this categorical feature to a numerical one.

Then, I split my data into a training data set and a test data set so I can be able to test how close the real data to the predicted one.

I tried to train different models:

- Linear Regression
- Random forest regressor
- K-neighbor regressor
- XGBoost Regressor

I plotted their r2\_score, which measure how well the regression line approximates the real data points.

I find that, in fact, I have for all the regressors a low r2\_score. And the xgboost have the best score in comparison with the other regressors.

The low score can be explained by the fact that we don't have enough features that participate in the definition of that salary and that train our model.

As I mentioned at first, I noticed a huge variability in the starting salaries and that doesn't necessarily increase or decrease with the hiring years. Maybe other features can be considered in the definition of the starting salary of each employee, like his diploma, his age, his previous experiences and achievements...

#### 4 - Validation :

Even if all the regressors don't have a good r2\_score, In the validation part I used the XGBoost regressor to calculate the starting salary of a senior engineer that will join the company in

August 2002, and his salary after 10 years in the company. I considered the both cases where the Senior Engineer is a man or a women.

And those were the results:

emp_no	gender	hire_year	hire_month	experience	Salary estimation
500000	M	2002	8	0	62219
500001	F	2002	8	0	60770
500000	M	2002	8	10	76434
500001	F	2002	8	10	75349

### How can we ameliorate the models:

We can tune the parameters of each model so to increase it performance or make it easier to train the model.