## Naukri.com job sample data analysis and visualization

Problem: Anticipate job market demand, (forecast)

solution : lets tackle this promblem , i.e forecasting the job market demand by analysis of job sample of naukri.com  $\,$ 

project:

## Major analysis we are going to do with the naukri.comjobsample data are :

- 1.Company's analysis
- 2.Industry's analysis
- 3. Jobtitle's analysis
- 4.Skill's analysis
- 5.Pay Rate's analysis i.e MIN and MAX Pay Rates and we will add this columns to our updated data
- 6.Experience analysis i.e MIN and MAX Experience and we will add this columns to our updated\_data
- 5.1 now the updated data contains 3 extra columns i.e minimum, maximum, average pay
- 6.1 now the updated data contains 3 extra columns i.e minimum, maximum, average experience
- 7. Experience and Pay\_Rates
- 7.1) co-relation between min\_experience and min\_pay
- a) seaborn stripplots
- b) seaborn pointplots
- 7.2) co-relation between max\_experience and max\_pay
- a) seaborn stripplots
- b) seaborn pointplots
- 8. MINIMUM, MAXIMUM EXPERIENCE AND MINIMUM, MAXIMUM PAY
- 8.1) co-relation between MIN\_experience, MAX\_experience and MIN\_pay
- 8.2) co-relation between MIN\_experience, MAX\_experience and MAX\_pay
- 9. co-relation between avg\_experience and avg\_pay
- 9.1) seaborn stripplots
- 9.2) seaborn pointplots

#### 10. comparison between

- 10.1) MINIMUM PAY & INDUSTRIES
- 10.2) MAXIMUM PAY & INDUSTRIES
- 10.3) AVERAGE PAY AND SKILLS
- 10.4) AVERAGE PAY AND JOBTITLES
- AA CHIRARAADW.

#### let's begin with importing the necessary libraries

In [2]:

```
import numpy as np #for algebric calculations
import pandas as pd #essential for data reading,writing etc
import seaborn as sns #visualization library
import matplotlib.pyplot as plt #visualization library
%matplotlib inline
plt.rcParams['figure.figsize'] = (10, 7) #plotting parameters size's
import warnings
warnings.filterwarnings('ignore')
```

#### After successfully importing libraries, now let's import the data

```
In [3]:
```

```
data = pd.read_csv("naukri_com-job_sample.csv")
```

let's read the first 10 columns of our imported data by using a pandas function head(10)

In [4]:

data.head(10)

#### Out[4]:

	company	education	experience	industry	jobdescription	
0	MM Media Pvt Ltd	UG: B.Tech/B.E Any Specialization PG:Any Po	0 - 1 yrs	Media / Entertainment / Internet	Job Description Send me Jobs like this Quali	210516
1	find live infotech	UG: B.Tech/B.E Any Specialization PG:MBA/PG	0 - 0 yrs	Advertising / PR / MR / Event Management	Job Description Send me Jobs like this Quali	210516
2	Softtech Career Infosystem Pvt. Ltd	UG: Any Graduate - Any Specialization PG:Any P	4 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this - as	101016
3	Onboard HRServices LLP	UG: Any Graduate - Any Specialization PG:CA Do	11 - 15 yrs	Banking / Financial Services / Broking	Job Description Send me Jobs like this - Inv	810169
4	Spire Technologies and Solutions Pvt. Ltd.	UG: B.Tech/B.E Any Specialization PG:Any Po	6 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this Pleas	120916
5	PFS Web Global Services Pvt Ltd	UG: B.Tech/B.E Any Specialization PG:MCA	2 - 5 yrs	IT-Software / Software Services	Job Description Send me Jobs like this We ar	131016
6	Kinesis Management Consultant Pvt. Ltd	NaN	1 - 3 yrs	IT-Software / Software Services	Job Description Send me Jobs like this exper	131016
7	Agile HR consultancy Pvt. Ltd. hiring for Ross	UG: Diploma - Any Specialization, Electrical,	2 - 7 yrs	Aviation / Aerospace Firms	Job Description Send me Jobs like this Job D	121016
8	HANSUM INDIA ELECTRONICS PVT.LTD.	UG: Diploma - Any Specialization, Electronics/	1 - 3 yrs	Industrial Products / Heavy Machinery	Job Description Send me Jobs like this Indep	131016

	company	education	experience	industry	jobdescription	
9	Accenture	UG: Any Graduate - Any Specialization	1 - 5 yrs	IT-Software / Software	Job Description Send me Jobs like this	121016

now we have an idea regarding the features we have in the data

let's begin the analysis, as we will get more in-depth useful insights of the data

#### let's see what are the top 10 companies?

#### In [5]:

בוו [כ].	
data['company'].value_counts().head(10)	
Out[5]:	
Indian Institute of Technology Bombay	403
Confidential	393
National Institute of Industrial Engineering	185
Oracle India Pvt. Ltd.	151
JPMorgan Chase	135
Godrej Industries Ltd	125
Unitforce technologies Pvt. Ltd.	100
Capgemini	98
HCL Technologies	95
Axis Jobs	92
Name: company, dtype: int64	

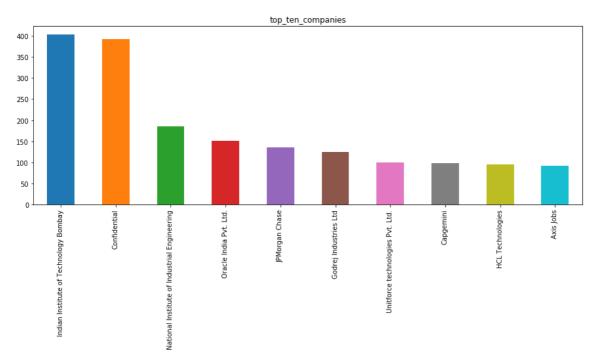
let's plot plot 10 companies for indepth analysis.

#### In [6]:

```
top_ten_companies=data['company'].value_counts().head(10)
f,ax=plt.subplots(figsize=(15,5))
top_ten_companies.plot(kind = 'bar')
plt.title('top_ten_companies')
```

#### Out[6]:

Text(0.5,1,'top\_ten\_companies')



it is clear that IITbombay is NO.1 among the top 10 company of our data

let's see what are the top 10 industries?

let's plot plot 10 industries for indepth insights.

#### In [7]:

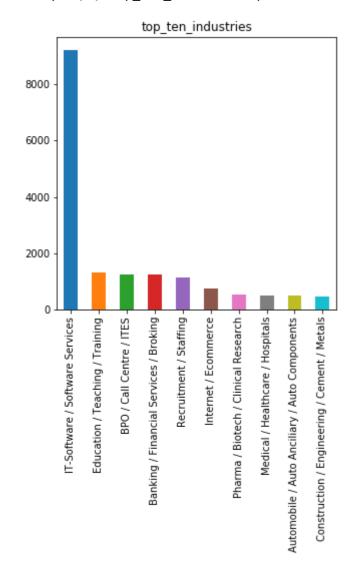
<pre>data['industry'].value_counts().head(10)</pre>	
Out[7]:	
IT-Software / Software Services	9216
Education / Teaching / Training	1322
BPO / Call Centre / ITES	1254
Banking / Financial Services / Broking	1238
Recruitment / Staffing	1129
Internet / Ecommerce	738
Pharma / Biotech / Clinical Research	525
Medical / Healthcare / Hospitals	495
Automobile / Auto Anciliary / Auto Components	478
Construction / Engineering / Cement / Metals Name: industry, dtype: int64	449

#### In [8]:

```
top_ten_industries = data['industry'].value_counts().head(10)
f,ax=plt.subplots(figsize=(5,5))
top_ten_industries.plot(kind='bar')
plt.title('top_ten_industries')
```

#### Out[8]:

Text(0.5,1,'top\_ten\_industries')



#### let's see what are the top 10 jobtitle's?

#### let's plot plot 10 jobtitle's for indepth insights.

#### In [9]:

data['jobtitle'].value\_counts().head(10)

#### Out[9]:

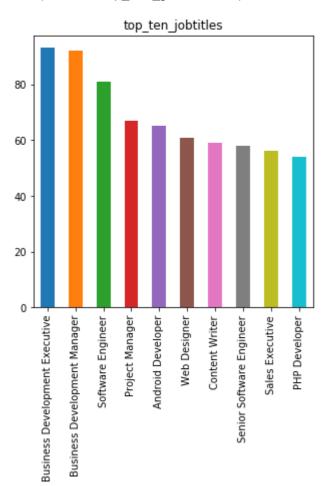
Business Development Executive	93
Business Development Manager	92
Software Engineer	81
Project Manager	67
Android Developer	65
Web Designer	61
Content Writer	59
Senior Software Engineer	58
Sales Executive	56
PHP Developer	54
Name: jobtitle, dtype: int64	

#### In [10]:

```
top_ten_jobtitles = data['jobtitle'].value_counts().head(10)
f,ax=plt.subplots(figsize=(5,5))
top_ten_jobtitles.plot(kind='bar')
plt.title('top_ten_jobtitles')
```

#### Out[10]:

Text(0.5,1,'top\_ten\_jobtitles')



#### let's see what are the top 10 skill's?

#### let's plot plot 10 skill's for indepth insights.

#### In [11]:

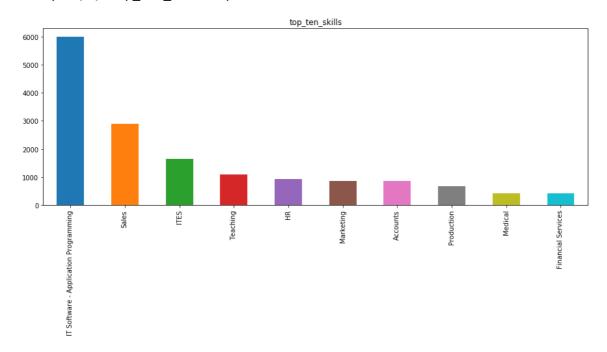
<pre>data['skills'].value_counts().head(10)</pre>	
Out[11]:	
IT Software - Application Programming	5989
Sales	2893
ITES	1640
Teaching	1091
HR	928
Marketing	868
Accounts	860
Production	667
Medical	418
Financial Services	413
Name: skills, dtype: int64	

#### In [12]:

```
top_ten_skills = data['skills'].value_counts().head(10)
f,ax=plt.subplots(figsize=(15,5))
top_ten_skills.plot(kind='bar')
plt.title('top_ten_skills')
```

#### Out[12]:

Text(0.5,1,'top\_ten\_skills')



on looking carefully the data given here contains the name of the old names of the cities along with some new names

this will cause error's in the analysis resulting in major mistakes

#### to avoid that , we will replace the old names with the new names and we will modify the data witgh the new names

#### In [13]:

```
replacements = {
   'joblocation_address': {
      r'(Bengaluru/Bangalore)': 'Bangalore',
      r'Bengaluru': 'Bangalore',
      r'Hyderabad / Secunderabad': 'Hyderabad',
      r'Mumbai , Mumbai': 'Mumbai',
      r'Noida': 'NCR',
      r'Delhi': 'NCR'
      r'Gurgaon': 'NCR',
      r'Delhi/NCR(National Capital Region)': 'NCR',
      r'Delhi , Delhi': 'NCR',
      r'Noida , Noida/Greater Noida': 'NCR'.
      r'Ghaziabad': 'NCR',
      r'Delhi/NCR(National Capital Region), Gurgaon': 'NCR',
      r'NCR , NCR': 'NCR',
      r'NCR/NCR(National Capital Region)': 'NCR',
      r'NCR , NCR/Greater NCR': 'NCR',
      r'NCR/NCR(National Capital Region), NCR': 'NCR',
      r'NCR , NCR/NCR(National Capital Region)': 'NCR',
      r'Bangalore , Bangalore': 'Bangalore',
      r'Bangalore , karnataka': 'Bangalore',
      r'NCR/NCR(National Capital Region)': 'NCR',
      r'NCR/Greater NCR': 'NCR',
      r'NCR/NCR(National Capital Region) , NCR': 'NCR'
   }
}
data.replace(replacements, regex=True, inplace=True)
y = data['joblocation_address'].value_counts()
```

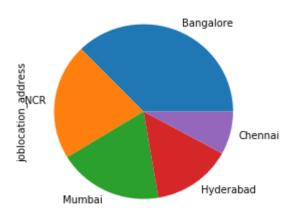
let's examine from where the more job's are coming?

#### In [14]:

```
most_job_posting_city=data['joblocation_address'].value_counts().head()
f ,ax=plt.subplots(figsize=(4,4))
most_job_posting_city.plot(kind = 'pie')
```

#### Out[14]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131c69f9f60>



it's clear now, that BENGALURU holds the first place in more joblocation.

thats a pretty good insights regading the

1.top10companies

2.top10industries

3.top10skills

4.jobtitles

5.top city for joblocations

lets examine the payrates:)

#### In [15]:

```
pay_split = data['payrate'].str[1:-1].str.split('-', expand=True)
pay_split.head()
```

#### Out[15]:

	0	1	2	3	4	5	6
0	,50,000	2,25,000 P.	None	None	None	None	None
1	,50,000	2,50,000 P.A. 2000	None	None	None	None	None
2	ot Disclosed by Recruite	None	None	None	None	None	None
3	ot Disclosed by Recruite	None	None	None	None	None	None
4	ot Disclosed by Recruite	None	None	None	None	None	None

#### In [16]:

```
#let's clean the payrates more
#remove space in left and right
pay_split[0] = pay_split[0].str.strip()
#remove comma
pay_split[0] = pay_split[0].str.replace(',', '')
#remove all character in two condition
# 1 remove if only character
# 2 if start in number remove after all character
pay_split[0] = pay_split[0].str.replace(r'\D.*', '')
#display
pay_split[0].head()
```

#### Out[16]:

```
0
     50000
1
     50000
2
3
```

Name: 0, dtype: object

#### In [17]:

```
#remove space in left and right
pay_split[1] = pay_split[1].str.strip()
#remove comma
pay_split[1] = pay_split[1].str.replace(',', '')
#remove all character in two condition
# 1 remove if only character
# 2 if start in number remove after all character
pay_split[1] = pay_split[1].str.replace(r'\D.*', '')
#display
pay_split[1].head()
```

#### Out[17]:

```
0
     225000
1
     250000
2
       None
```

3 None

None

Name: 1, dtype: object

#### In [18]:

```
pay_split[0] = pd.to_numeric(pay_split[0], errors='coerce')
pay_split[1] = pd.to_numeric(pay_split[1], errors='coerce')
```

#### that's the end of massive cleaning part , now its time to concatenate the new results

```
In [19]:
```

```
pay=pd.concat([pay_split[0], pay_split[1]], axis=1, sort=False)
```

#### hurrey! we done it , lets name the new borns:)

```
In [20]:
```

```
pay.rename(columns={0:'min_pay', 1:'max_pay'}, inplace=True)
pay.head()
```

#### Out[20]:

	min_pay	max_pay
0	50000.0	225000.0
1	50000.0	250000.0
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN

#### the big work, now lets add this new bornone's to the data

#### In [21]:

```
#sometimes while testing , we may get multiple min_pay, max_pay columns as we are using
 conaction function for previous data to new data
#in such cases , we can use following code for removal of extra min-pay, max_pay columns
#data = data.drop("min_pay",axis = 1)
#data = data.drop("max_pay",axis = 1)
#data.head()
data=pd.concat([data, pay], axis=1, sort=False)
data.head()
```

#### Out[21]:

	company	education	experience	industry	jobdescription	job
0	MM Media Pvt Ltd	UG: B.Tech/B.E Any Specialization PG:Any Po	0 - 1 yrs	Media / Entertainment / Internet	Job Description Send me Jobs like this Quali	2105160022€
1	find live infotech	UG: B.Tech/B.E Any Specialization PG:MBA/PG	0 - 0 yrs	Advertising / PR / MR / Event Management	Job Description Send me Jobs like this Quali	2105160023\$
2	Softtech Career Infosystem Pvt. Ltd	UG: Any Graduate - Any Specialization PG:Any P	4 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this - as	10101690053
3	Onboard HRServices LLP	UG: Any Graduate - Any Specialization PG:CA Do	11 - 15 yrs	Banking / Financial Services / Broking	Job Description Send me Jobs like this - Inv	8101690053€
4	Spire Technologies and Solutions Pvt. Ltd.	UG: B.Tech/B.E Any Specialization PG:Any Po	6 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this Pleas	12091600212

#### In [22]:

```
experience_split = data['experience'].str[0:-1].str.split('-', expand=True)
experience_split.head()
```

#### Out[22]:

	0	1	2
0	0	1 yr	None
1	0	0 yr	None
2	4	8 yr	None
3	11	15 yr	None
4	6	8 yr	None

#### In [23]:

```
#let's clean the experience more
#remove space in left and right
experience_split[1] = experience_split[1].str.strip()
#remove comma
experience_split[1] = experience_split[1].str.replace('yr', '')
#remove all character in two condition
# 1 remove if only character
# 2 if start in number remove after all character
experience_split[1] = experience_split[1].str.replace(r'yr', '')
#display
experience_split[1].head()
```

#### Out[23]:

```
0
        1
```

2 8

3 15

Name: 1, dtype: object

#### In [24]:

```
experience split[0] = pd.to numeric(experience split[0], errors='coerce')
experience split[1] = pd.to numeric(experience split[1], errors='coerce')
```

#### In [25]:

```
experience=pd.concat([experience split[0], experience split[1]], axis=1, sort=False)
```

<sup>1</sup> 0

#### In [26]:

experience.rename(columns={0:'min\_experience', 1:'max\_experience'}, inplace=True) experience.head()

#### Out[26]:

	min_experience	max_experience
0	0.0	1.0
1	0.0	0.0
2	4.0	8.0
3	11.0	15.0
4	6.0	8.0

#### In [27]:

data=pd.concat([data, experience], axis=1, sort=False) data.head()

Out[27]:

	company	education	experience	industry	jobdescription	job
0	MM Media Pvt Ltd	UG: B.Tech/B.E Any Specialization PG:Any Po	0 - 1 yrs	Media / Entertainment / Internet	Job Description Send me Jobs like this Quali	21051600226
1	find live infotech	UG: B.Tech/B.E Any Specialization PG:MBA/PG	0 - 0 yrs	Advertising / PR / MR / Event Management	Job Description Send me Jobs like this Quali	21051600239
2	Softtech Career Infosystem Pvt. Ltd	UG: Any Graduate - Any Specialization PG:Any P	4 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this - as	10101690053
3	Onboard HRServices LLP	UG: Any Graduate - Any Specialization PG:CA Do	11 - 15 yrs	Banking / Financial Services / Broking	Job Description Send me Jobs like this - Inv	81016900536
4	Spire Technologies and Solutions Pvt. Ltd.	UG: B.Tech/B.E Any Specialization PG:Any Po	6 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this Pleas	12091600212

### 5.1 & 6.1

Now it's time to add some average values of payrate's and experience , giving us more deeper insights

#### In [28]:

```
data['avg_pay']=(data['min_pay'].values + data['max_pay'].values)/2
data['avg_experience']=(data['min_experience'].values + data['max_experience'].values)/
data.head()
```

Out[28]:

	company	education	experience	industry	jobdescription	job
0	MM Media Pvt Ltd	UG: B.Tech/B.E Any Specialization PG:Any Po	0 - 1 yrs	Media / Entertainment / Internet	Job Description Send me Jobs like this Quali	2105160022€
1	find live infotech	UG: B.Tech/B.E Any Specialization PG:MBA/PG	0 - 0 yrs	Advertising / PR / MR / Event Management	Job Description Send me Jobs like this Quali	2105160023\$
2	Softtech Career Infosystem Pvt. Ltd	UG: Any Graduate - Any Specialization PG:Any P	4 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this - as	10101690053
3	Onboard HRServices LLP	UG: Any Graduate - Any Specialization PG:CA Do	11 - 15 yrs	Banking / Financial Services / Broking	Job Description Send me Jobs like this - Inv	8101690053€
4	Spire Technologies and Solutions Pvt. Ltd.	UG: B.Tech/B.E Any Specialization PG:Any Po	6 - 8 yrs	IT-Software / Software Services	Job Description Send me Jobs like this Pleas	12091600212

## 7. Experience and Pay\_Rates

Relation between experience and pay\_rates

this will give us an idea how your experience impacts your pay\_rates

7.1) co-relation between min\_experience and min\_pay

let's get the insights by using

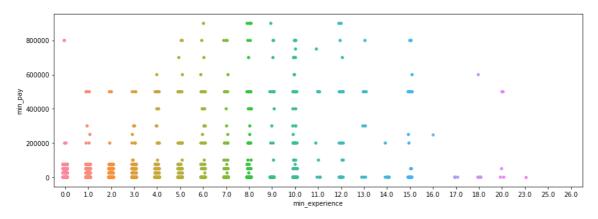
#### A)seaborn stripplots

#### In [29]:

```
f,ax=plt.subplots(figsize=(15,5))
sns.stripplot(x='min_experience', y='min_pay', data=data, jitter=True)
```

#### Out[29]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131c74443c8>



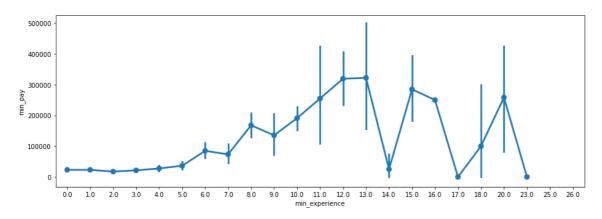
#### B)seaborn pointplots

#### In [30]:

```
f,ax=plt.subplots(figsize=(15,5))
sns.pointplot(x='min_experience', y='min_pay', data=data)
```

#### Out[30]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131c723ae48>



#### 7.2) co-relation between max\_experience and max\_pay

#### let's get the insights by using

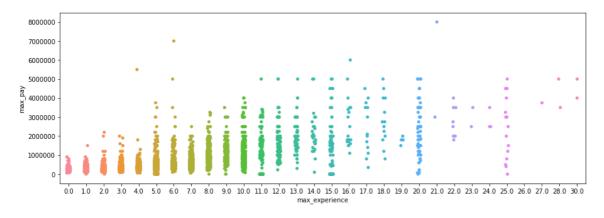
#### A)seaborn stripplots

#### In [31]:

```
f,ax=plt.subplots(figsize=(15,5))
sns.stripplot(x='max_experience', y='max_pay', data=data, jitter=True)
```

#### Out[31]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131c7899400>



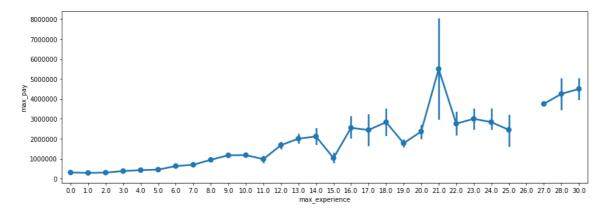
#### B) seaborn pointplots

#### In [32]:

```
f,ax=plt.subplots(figsize=(15,5))
sns.pointplot(x='max_experience', y='max_pay', data=data)
```

#### Out[32]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131ca980908>



## 8) MINIMUM, MAXIMUM experience and MINIMUM, MAXIMUM pay

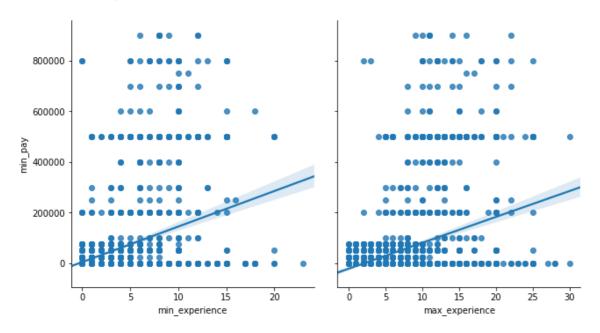
## 8.1) Relation between MIN, MAX experience and MIN pay

#### In [33]:

```
sns.pairplot(data,
             size=5, aspect=0.9,
             x_vars=["min_experience","max_experience"],
             y_vars=["min_pay"],
             kind="reg")
```

#### Out[33]:

<seaborn.axisgrid.PairGrid at 0x131c78f4cc0>



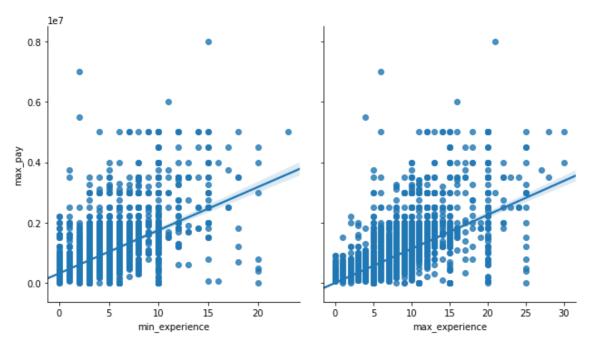
## 8.2) Relation between MIN, MAX experience and MAX pay

#### In [34]:

```
sns.pairplot(data,
             size=5, aspect=0.9,
             x_vars=["min_experience","max_experience"],
             y_vars=["max_pay"],
             kind="reg")
```

#### Out[34]:

<seaborn.axisgrid.PairGrid at 0x131c757c7b8>



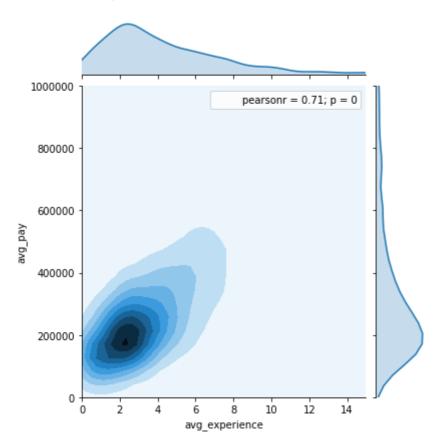
# 9) Relation between AVERAGE experience and AVERAGE pay\_rates

#### In [35]:

```
sns.jointplot(x='avg_experience', y='avg_pay', data=data,
              kind="kde",xlim={0,15}, ylim={0,1000000})
```

#### Out[35]:

<seaborn.axisgrid.JointGrid at 0x131c731bcc0>



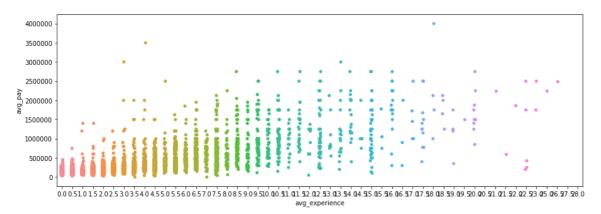
#### 9.1) co-relation between avg\_experience and avg\_pay by using seaborn.stripplots

#### In [36]:

```
f,ax=plt.subplots(figsize=(15,5))
sns.stripplot(x='avg_experience', y='avg_pay', data=data, jitter=True)
```

#### Out[36]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131c6e0d3c8>



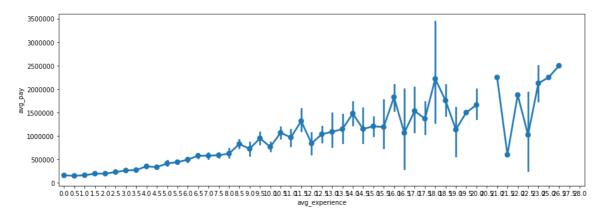
#### 9.2) co-relation between avg\_experience and avg\_pay by using seaborn.pointplots

#### In [37]:

```
f,ax=plt.subplots(figsize=(15,5))
sns.pointplot(x='avg_experience', y='avg_pay', data=data)
```

#### Out[37]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131cace8588>



#### 10. comparison between

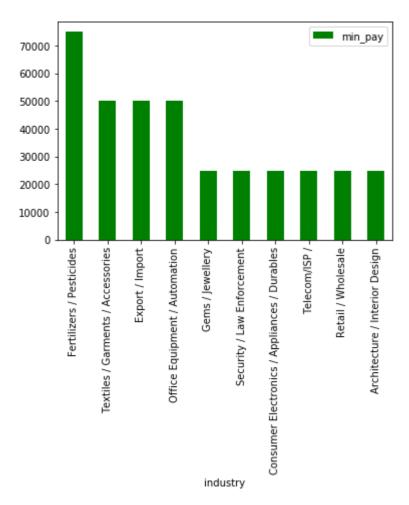
#### 10.1) MINIMUM PAY & INDUSTRIES

#### In [38]:

data[['min\_pay','industry']].groupby(["industry"]).median().sort\_values(by='min\_pay',as cending=False).head(10).plot.bar(color='green')

#### Out[38]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131cadfc860>



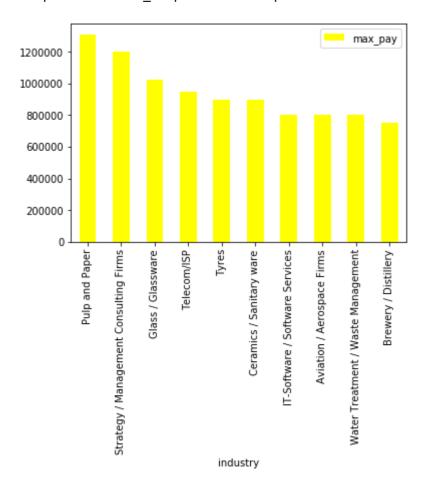
#### 10.2) MAXIMUM PAY & INDUSTRIES

#### In [39]:

data[['max\_pay','industry']].groupby(["industry"]).median().sort\_values(by='max\_pay',as cending=False).head(10).plot.bar(color='yellow')

#### Out[39]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131cc1a6a58>



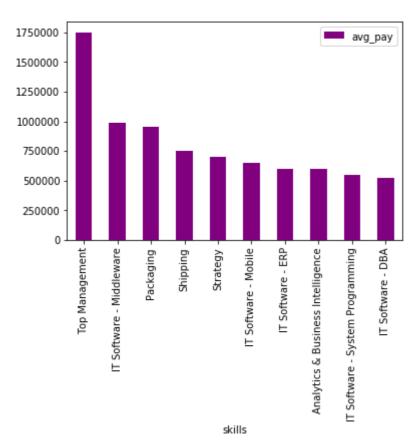
#### 10.3) AVERAGE PAY AND SKILLS

#### In [40]:

data[['avg\_pay','skills']].groupby(["skills"]).median().sort\_values(by='avg\_pay',ascend ing=False).head(10).plot.bar(color='purple')

#### Out[40]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131cc22c7b8>



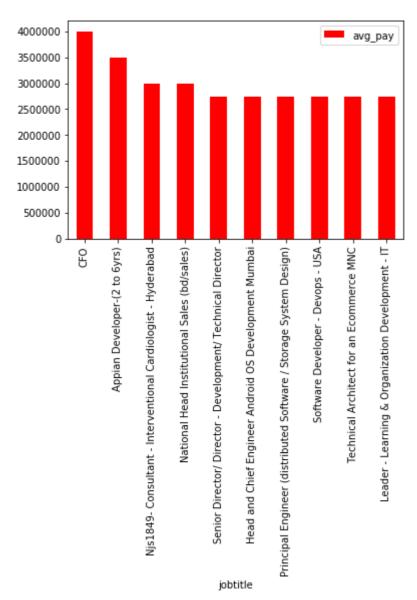
#### 10.4) AVERAGE PAY AND JOBTITLES

#### In [41]:

data[['avg\_pay','jobtitle']].groupby(["jobtitle"]).median().sort\_values(by='avg\_pay',as cending=False).head(10).plot.bar(color='r')

#### Out[41]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x131cc2ae240>



### Result

1. we majorly speak here about the expereince's and pay\_rates.

2. secondly,	we will s	peak about	the coma	parison	between	the
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a) industries and pay\_rates LA ---- ---- --- d -1-!!!- !-|-4!41--

Thanking you , satyamsharma(Highly Passionate Machine Learning Engineer)