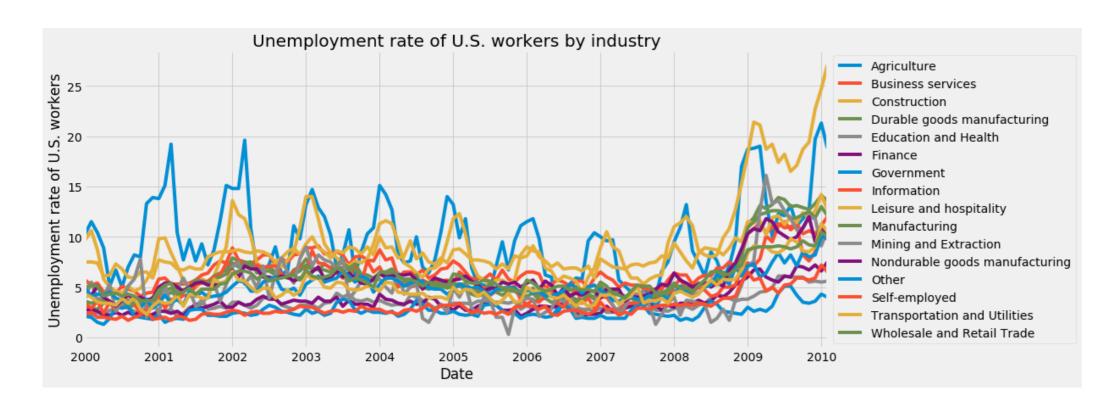
# Apply your knowledge to a new dataset

VISUALIZING TIME SERIES DATA IN PYTHON





#### The Jobs dataset



### Let's get started!



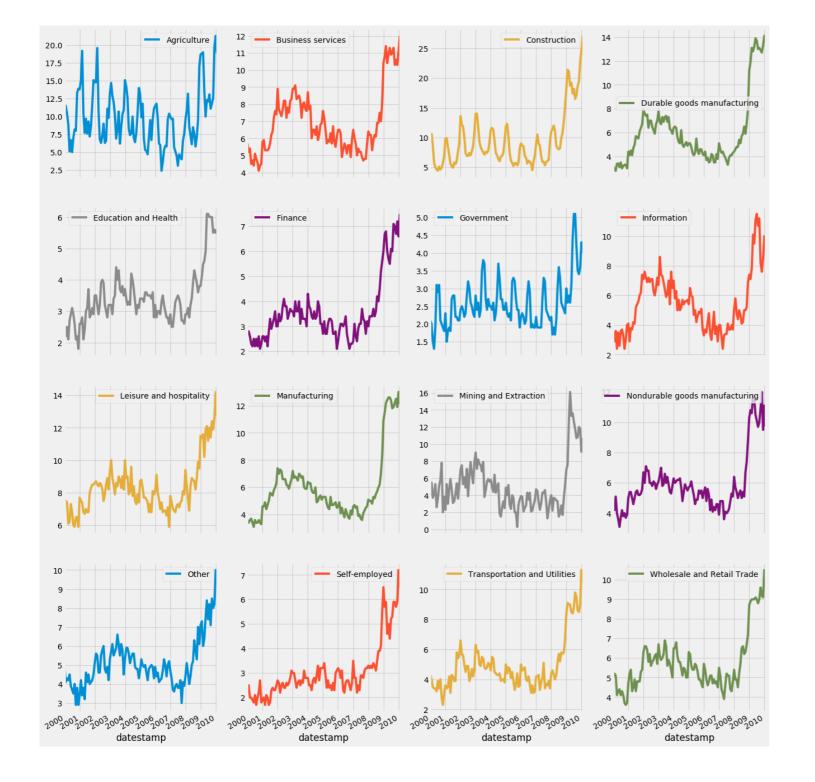
## Beyond summary statistics

VISUALIZING TIME SERIES DATA IN PYTHON

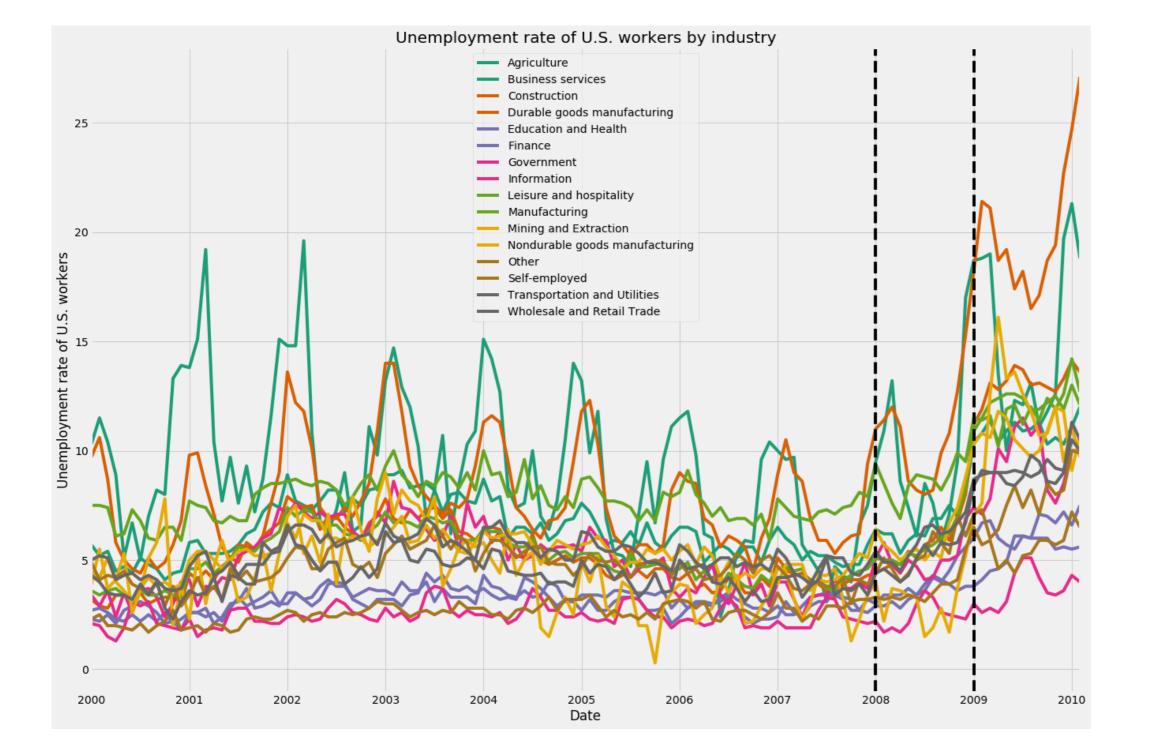




### Facet plots of the jobs dataset



### Annotating events in the jobs dataset



### Taking seasonal average in the jobs dataset

```
print(jobs.index)
```

```
DatetimeIndex(['2000-01-01', '2000-02-01', '2000-03-01', '2000-04-01', '2009-09-01', '2009-10-01', '2009-11-01', '2009-12-01', '2010-01-01', '2010-02-01'], dtype='datetime64[ns]', name='datestamp', length=122, freq=None)
```

```
index_month = jobs.index.month
jobs_by_month = jobs.groupby(index_month).mean()
print(jobs_by_month)
```

```
datestamp
            Agriculture
                         Business services
                                            Construction
                                7.863636
                                             12.909091
            13.763636
                                7.645455
            13.645455
                                             13.600000
            13.830000
                                7.130000
                                             11.290000
             9.130000
                                6.270000
                                              9.450000
             7.100000
                                6.600000
                                              8.120000
```

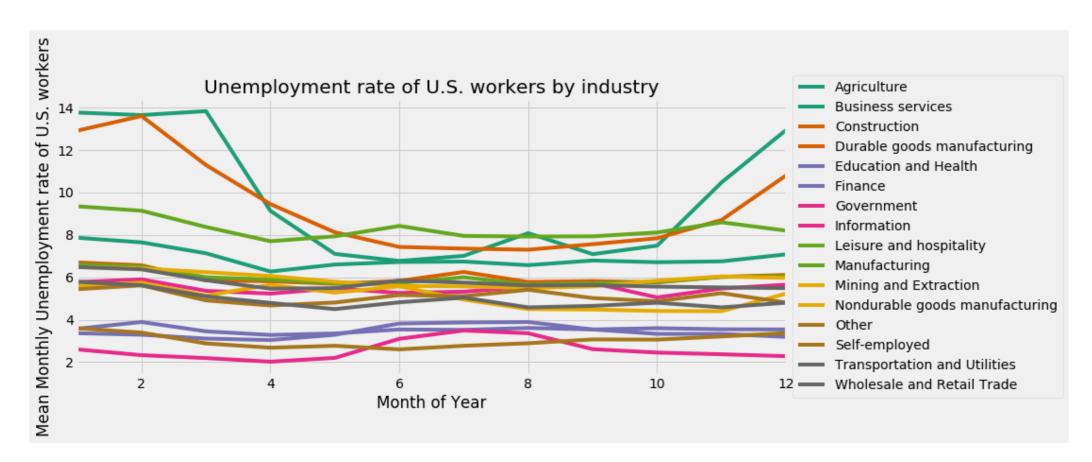


### Monthly averages in the jobs dataset

```
ax = jobs_by_month.plot(figsize=(12, 5),
colormap='Dark2')

ax.legend(bbox_to_anchor=(1.0, 0.5),
loc='center left')
```

### Monthly averages in the jobs dataset



### Time to practice!



## Decompose time series data

VISUALIZING TIME SERIES DATA IN PYTHON





### Python dictionaries

```
# Initialize a Python dictionnary
my_dict = {}
# Add a key and value to your dictionnary
my_dict['your_key'] = 'your_value'
# Add a second key and value to your dictionnary
my_dict['your_second_key'] = 'your_second_value'
# Print out your dictionnary
print(my_dict)
```

```
{'your_key': 'your_value',
'your_second_key': 'your_second_value'}
```



### Decomposing multiple time series with Python dictionaries

```
# Import the statsmodel library
import statsmodels.api as sm
# Initialize a dictionary
my_dict = {}
# Extract the names of the time series
ts_names = df.columns
print(ts_names)
```

```
['ts1', 'ts2', 'ts3']
```

```
# Run time series decomposition
for ts in ts_names:
    ts_decomposition = sm.tsa.seasonal_decompose(jobs[ts])
    my_dict[ts] = ts_decomposition
```

### Extract decomposition components of multiple time series

```
# Initialize a new dictionnary
my_dict_trend = {}
# Extract the trend component
for ts in ts_names:
    my_dict_trend[ts] = my_dict[ts].trend
# Convert to a DataFrame
trend_df = pd.DataFrame.from_dict(my_dict_trend)
print(trend_df)
```

```
ts1 ts2 ts3
datestamp
2000-01-01 2.2 1.3 3.6
2000-02-01 3.4 2.1 4.7
...
```

# Python dictionaries for the win!



# Compute correlations between time series

VISUALIZING TIME SERIES DATA IN PYTHON





#### Trends in Jobs data

print(trend\_df)

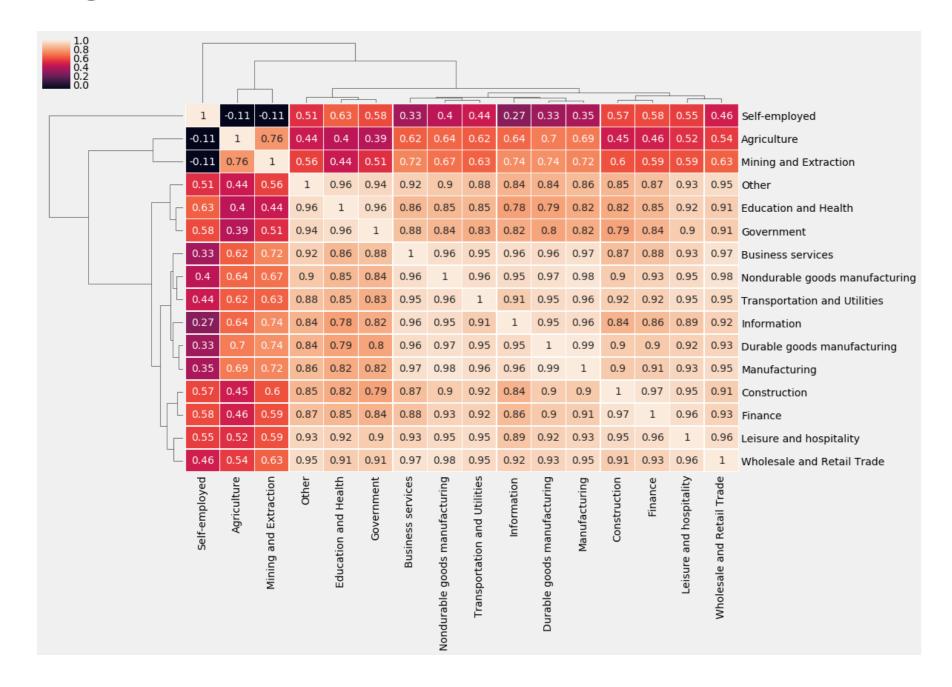
datestamp	Agriculture	Business services	Construction
2000-01-01	NaN	NaN	NaN
2000-02-01	NaN	NaN	NaN
2000-03-01	NaN	NaN	NaN
2000-04-01	NaN	NaN	NaN
2000-05-01	NaN	NaN	NaN
2000-06-01	NaN	NaN	NaN
2000-07-01	9.170833	4.787500	6.329167
2000-08-01	9.466667	4.820833	6.304167



### Plotting a clustermap of the jobs correlation matrix

```
# Get correlation matrix of the seasonality_df DataFrame
trend_corr = trend_df.corr(method='spearman')
# Customize the clustermap of the seasonality_corr
correlation matrix
fig = sns.clustermap(trend_corr, annot=True, linewidth=0.4)
plt.setp(fig.ax_heatmap.yaxis.get_majorticklabels(),
rotation=0)
plt.setp(fig.ax_heatmap.xaxis.get_majorticklabels(),
rotation=90)
```

### The jobs correlation matrix



### Let's practice!



### Congratulations!

VISUALIZING TIME SERIES DATA IN PYTHON





### Going further with time series

- Data from Zillow Research
- Kaggle competitions
- Reddit Data

### Going further with time series

- The importance of time series in business:
  - to identify seasonal patterns and trends
  - to study past behaviors
  - to produce robust forecasts
  - to evaluate and compare company achievements

### Getting to the next level

- Manipulating Time Series Data in Python
- Importing & Managing Financial Data in Python
- Statistical Thinking in Python (Part 1)
- Supervised Learning with scikit-learn

### Thank you!

