



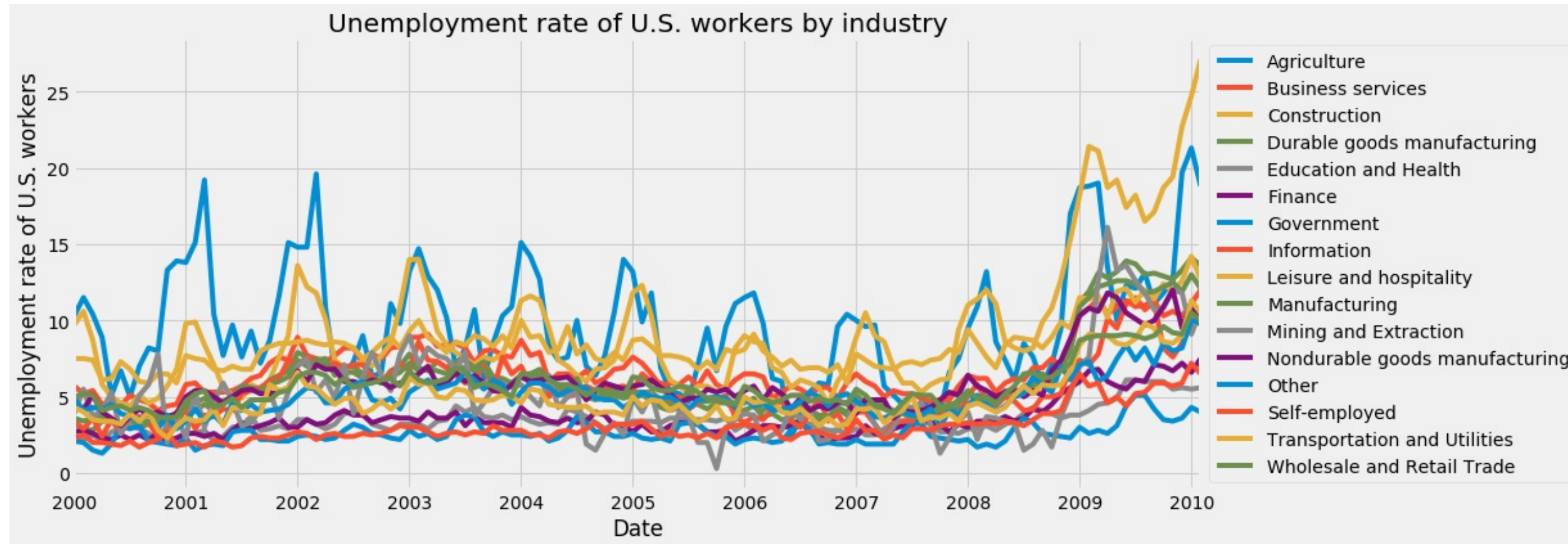
## VISUALIZING TIME SERIES DATA IN PYTHON

**Apply your knowledge  
to a new dataset**

Thomas Vincent

Head of Data Science, Getty Images

# The Jobs dataset





## VISUALIZING TIME SERIES DATA IN PYTHON

**Let's get started!**



VISUALIZING TIME SERIES DATA IN PYTHON

# Beyond summary statistics

Thomas Vincent

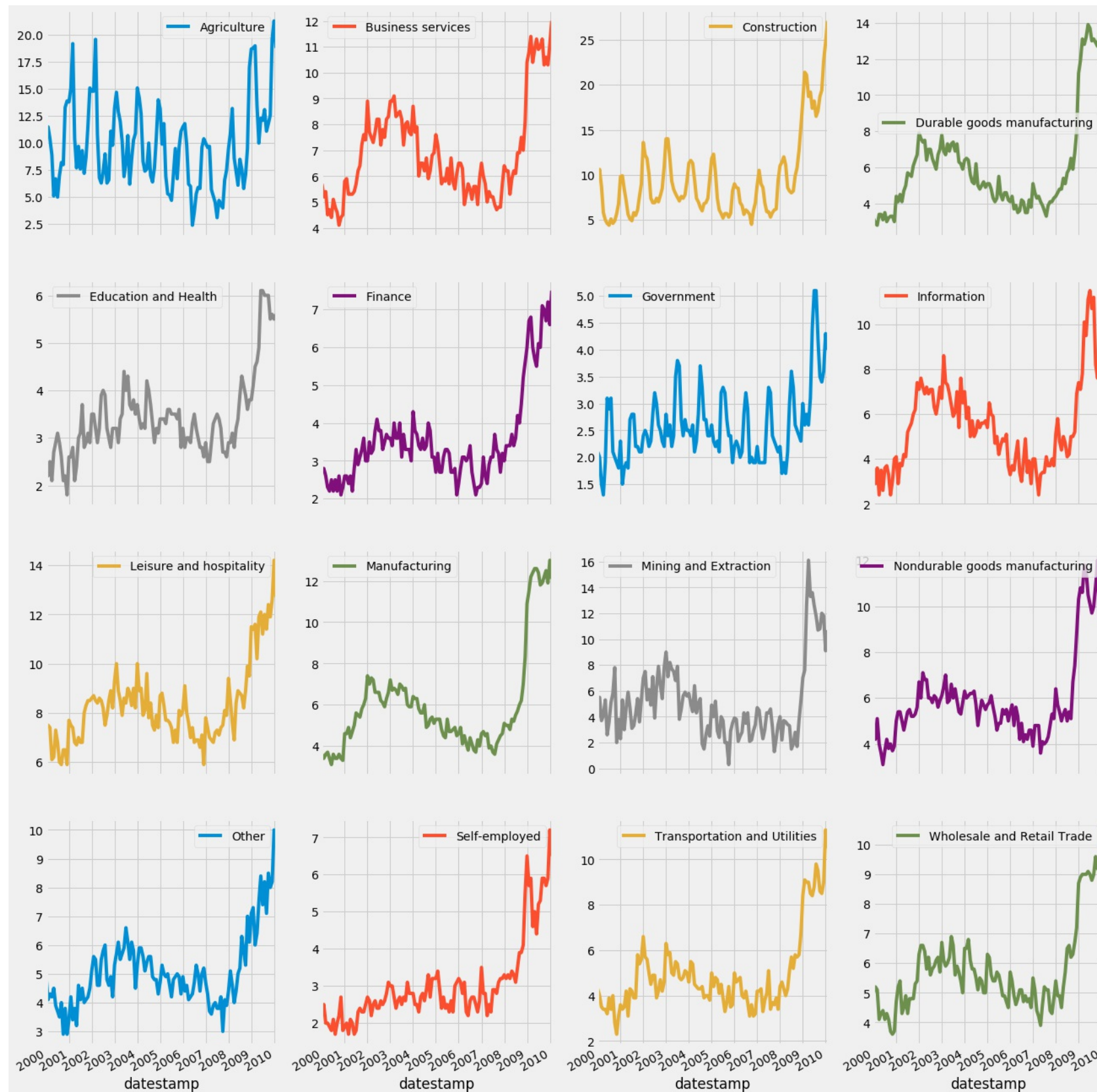
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# Facet plots of the jobs dataset

```
In [1]: jobs.plot(subplots=True,  
                  layout=(4, 4),  
                  figsize=(20, 16),  
                  sharex=True,  
                  sharey=False)
```

```
In [2]: plt.show()
```





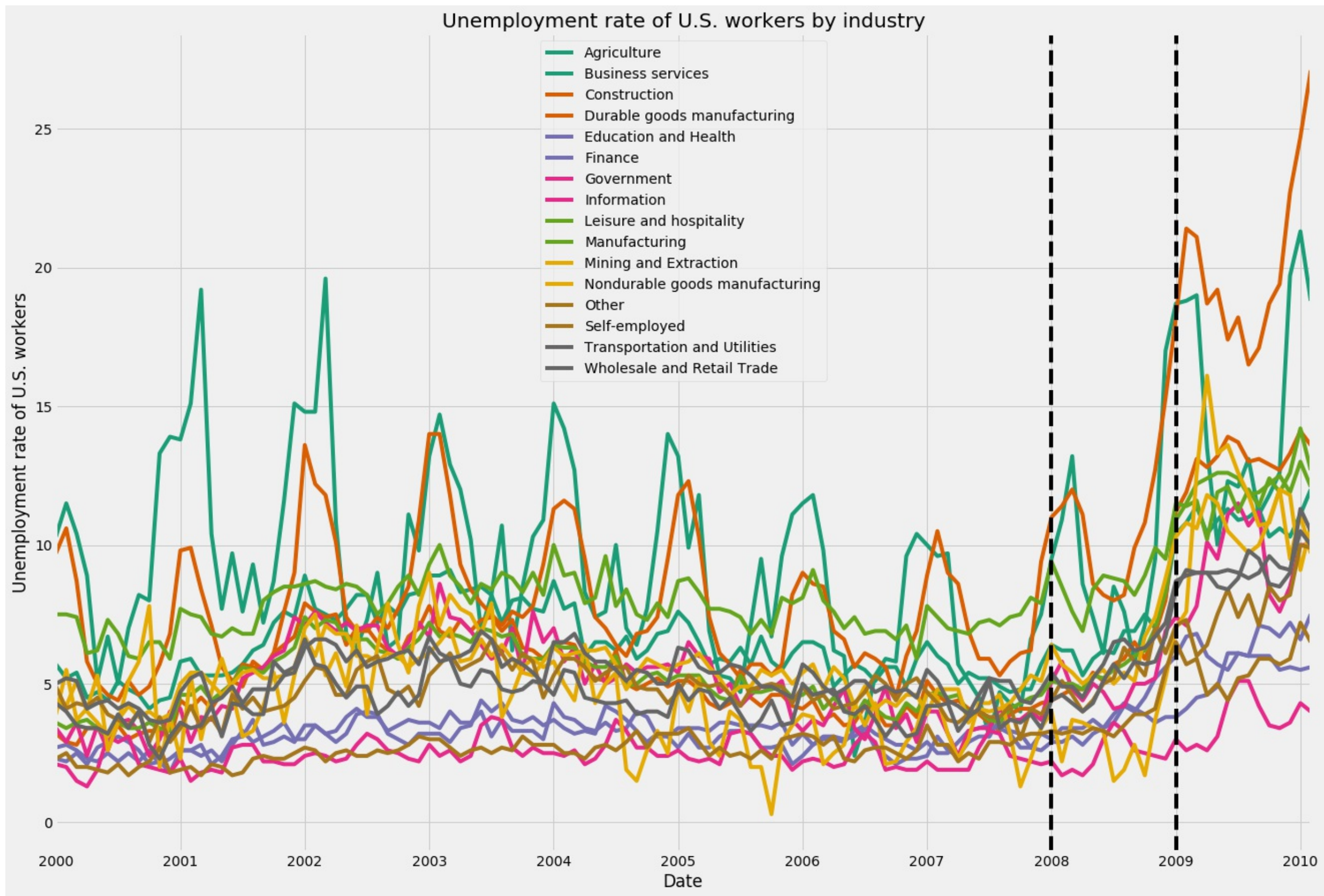
# Annotating events in the jobs dataset

```
In [1]: ax = jobs.plot(figsize=(20, 14), colormap='Dark2')
```

```
In [2]: ax.axvline('2008-01-01', color='black', linestyle='--')
```

```
In [3]: ax.axvline('2009-01-01', color='black', linestyle='--')
```









# Taking seasonal average in the jobs dataset

```
In [1]: 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)
```

```
In [2]: index_month = jobs.index.month
```

```
In [3]: jobs_by_month = jobs.groupby(index_month).mean()
```

```
In [4]: print(jobs_by_month)
           Agriculture  Business services  Construction \
datestamp
1           13.763636           7.863636           12.909091
2           13.645455           7.645455           13.600000
3           13.830000           7.130000           11.290000
4            9.130000           6.270000            9.450000
5            7.100000           6.600000            8.120000
...
```



# Monthly averages in the jobs dataset

```
In [1]: ax = jobs_by_month.plot(figsize=(12, 5), colormap='Dark2')
```

```
In [2]: ax.legend(bbox_to_anchor=(1.0, 0.5), loc='center left')
```





## VISUALIZING TIME SERIES DATA IN PYTHON

**Time to practice!**



VISUALIZING TIME SERIES DATA IN PYTHON

# Decompose time series data

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# Python dictionaries

```
# Initialize a Python dictionary
In [1]: my_dict = {}

# Add a key and value to your dictionary
In [2]: my_dict['your_key'] = 'your_value'

# Add a second key and value to your dictionary
In [3]: my_dict['your_second_key'] = 'your_second_value'

# Print out your dictionary
In [4]: print(my_dict)
{'your_key': 'your_value',
 'your_second_key': 'your_second_value'}
```



# Decomposing multiple time series with Python

## dictionaries

```
# Import the statsmodel library
In [1]: import statsmodels.api as sm

# Initialize a dictionary
In [2]: my_dict = {}

# Extract the names of the time series
In [3]: ts_names = df.columns

In [4]: print(ts_names)
['ts1', 'ts2', 'ts3']

# Run time series decomposition
In [5]: 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 dictionary
In [1]: my_dict_trend = {}

# Extract the trend component
In [2]: for ts in ts_names:
         my_dict_trend[ts] = my_dict[ts].trend

# Convert to a DataFrame
In [3]: trend_df = pd.DataFrame.from_dict(my_dict_trend)

In [4]: 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
...
```



VISUALIZING TIME SERIES DATA IN PYTHON

**Python dictionaries for  
the win!**



VISUALIZING TIME SERIES DATA IN PYTHON

# Compute correlations between time series

Thomas Vincent

Head of Data Science, Getty Images





# Trends in Jobs data

```
In [1]: print(trend_df)
          Agriculture  Business services  Construction \
datestamp
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

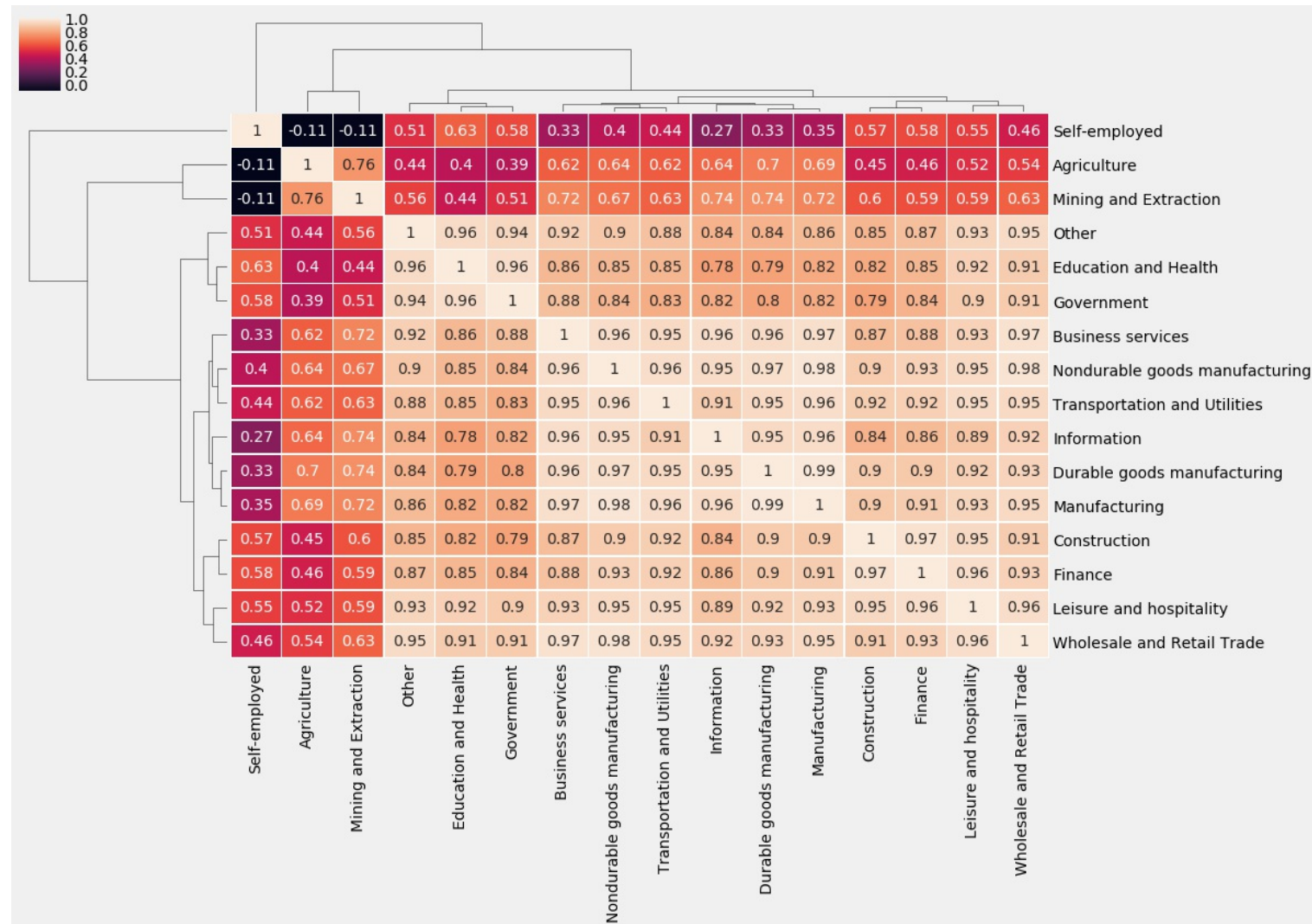
```
# Get correlation matrix of the seasonality_df DataFrame
In [1]: trend_corr = trend_df.corr(method='spearman')

# Customize the clustermap of the seasonality_corr correlation matrix
In [2]: fig = sns.clustermap(trend_corr, annot=True, linewidth=0.4)

In [3]: plt.setp(fig.ax_heatmap.yaxis.get_majorticklabels(), rotation=0)

In [4]: plt.setp(fig.ax_heatmap.xaxis.get_majorticklabels(), rotation=90)
```

# The jobs correlation matrix





## VISUALIZING TIME SERIES DATA IN PYTHON

# Congratulations!

Thomas Vincent

Head of Data Science, Getty Images



# 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 behaviours
  - 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](#)



## VISUALIZING TIME SERIES DATA IN PYTHON

**Thank you!**