

Apply your knowledge to a new dataset

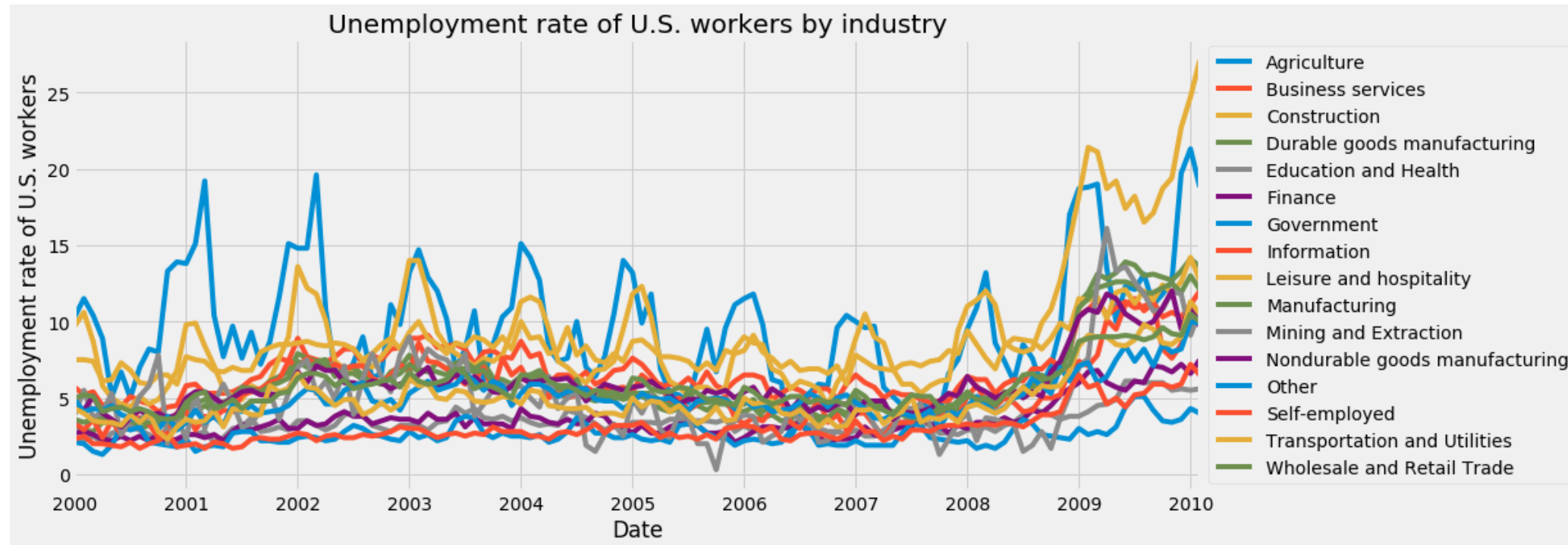
VISUALIZING TIME SERIES DATA IN PYTHON



Thomas Vincent

Head of Data Science, Getty Images

The Jobs dataset



Let's get started!

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Beyond summary statistics

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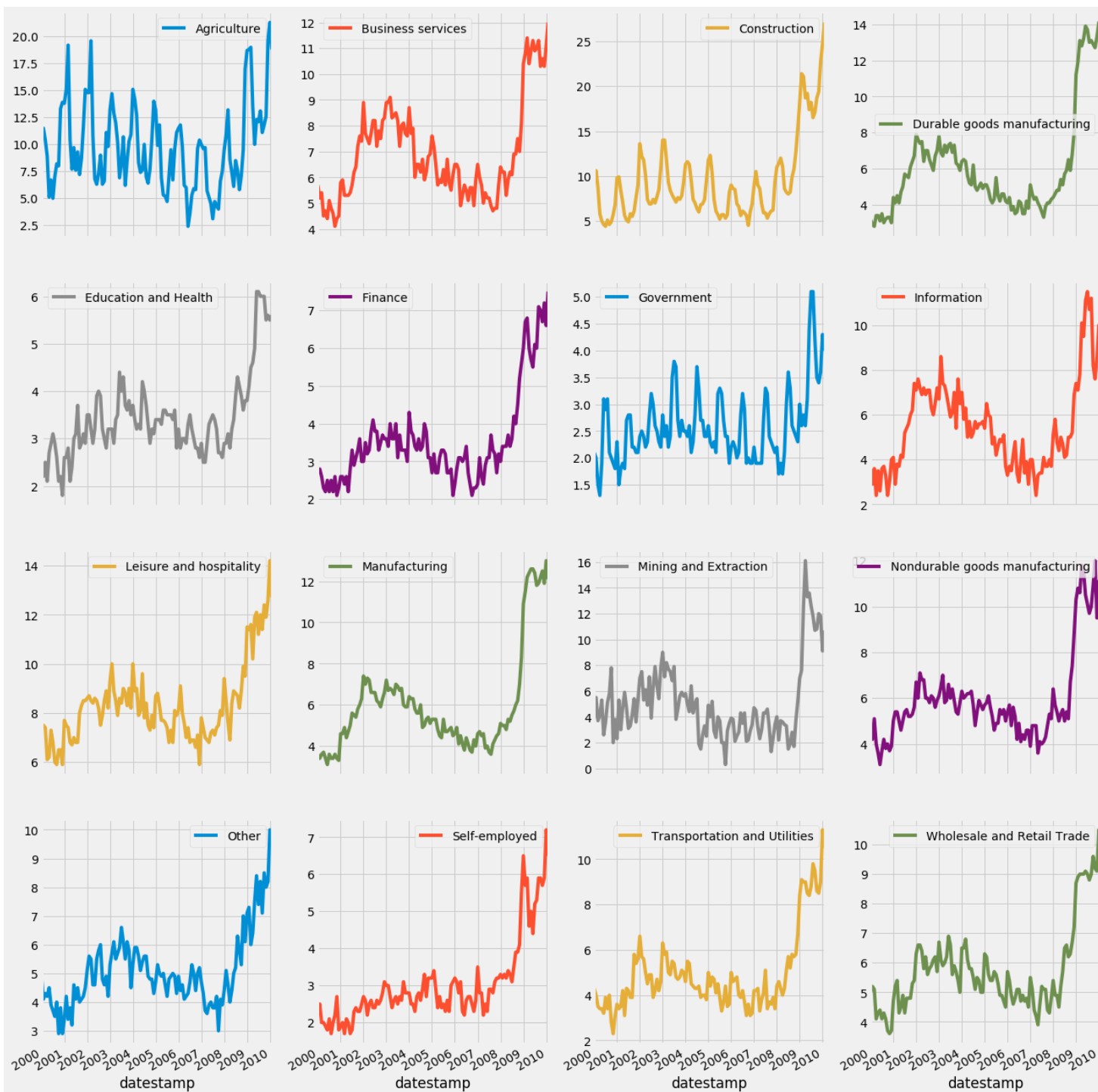


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Facet plots of the jobs dataset

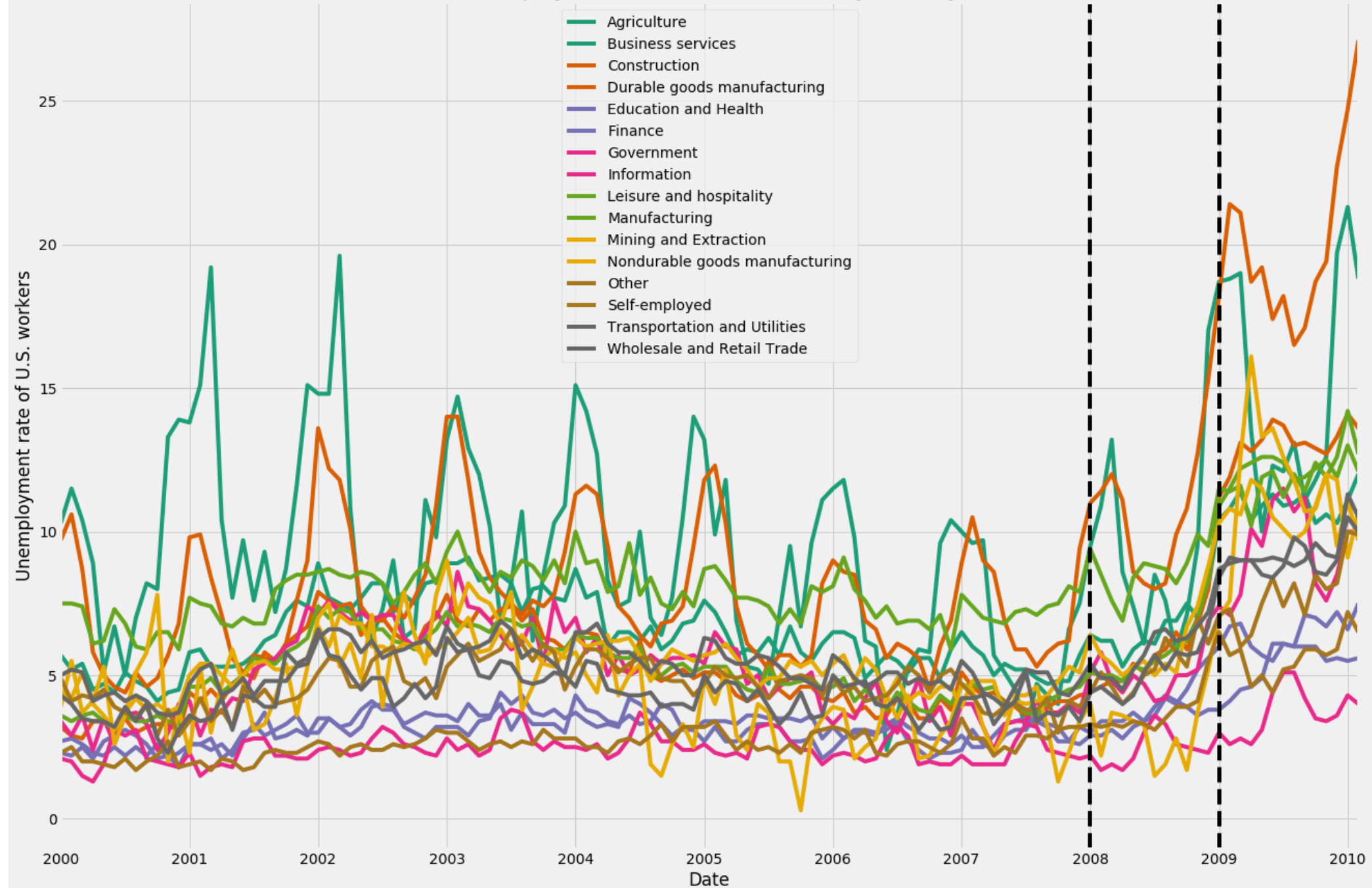
```
jobs.plot(subplots=True,  
          layout=(4, 4),  
          figsize=(20, 16),  
          sharex=True,  
          sharey=False)  
  
plt.show()
```



Annotating events in the jobs dataset

```
ax = jobs.plot(figsize=(20, 14), colormap='Dark2')
ax.axvline('2008-01-01', color='black',
           linestyle='--')
ax.axvline('2009-01-01', color='black',
           linestyle='--')
```

Unemployment rate of U.S. workers by industry



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)
```

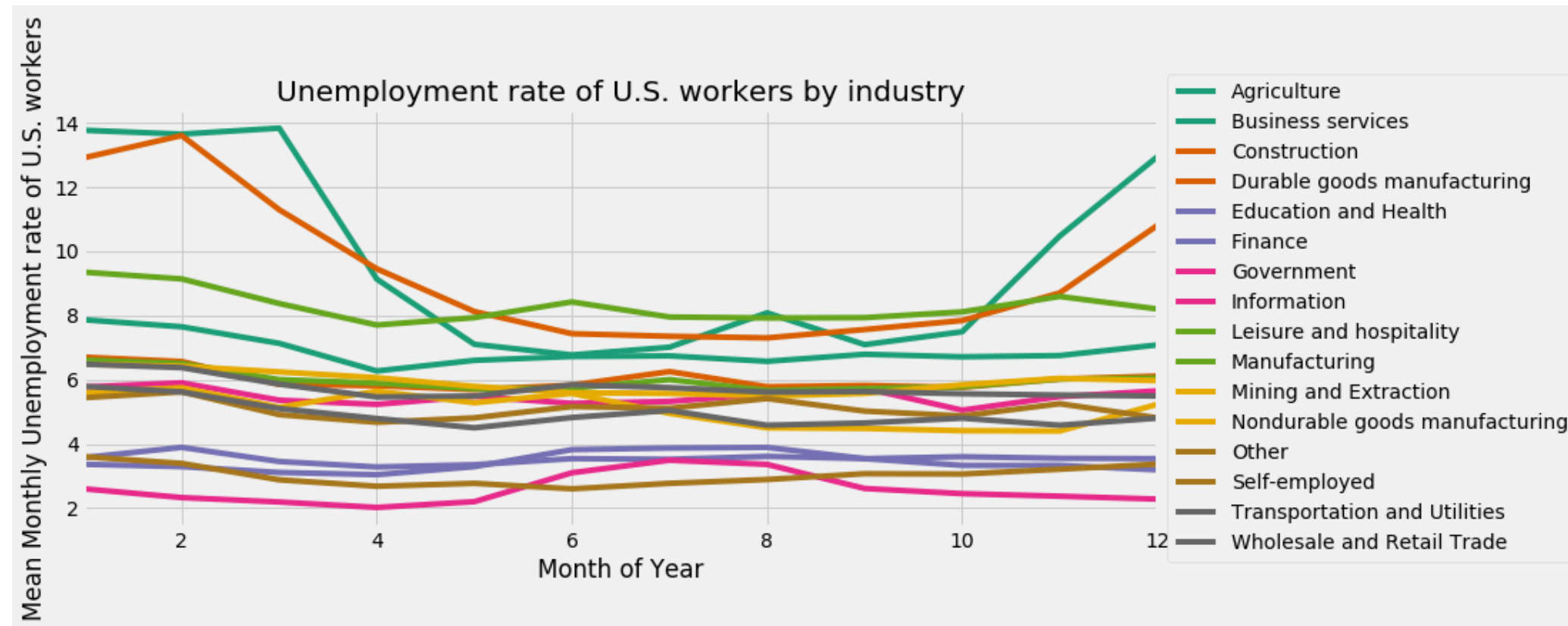
```
datestamp  Agriculture  Business services  Construction  
1          13.763636          7.863636      12.909091  
2          13.645455          7.645455      13.600000  
3          13.830000          7.130000      11.290000  
4           9.130000          6.270000       9.450000  
5           7.100000          6.600000       8.120000  
...
```

Monthly averages in the jobs dataset

```
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

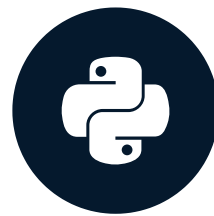


Time to practice!

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Decompose time series data

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Python dictionaries

```
# Initialize a Python dictionary
my_dict = {}

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

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

# Print out your dictionary
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 dictionary
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)
```

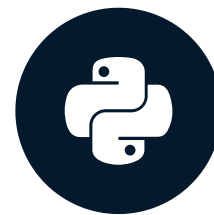
```
      ts1  ts2  ts3
datestamp
2000-01-01  2.2  1.3  3.6
2000-02-01  3.4  2.1  4.7
...
```


Python dictionaries for the win!

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Compute correlations between time series

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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

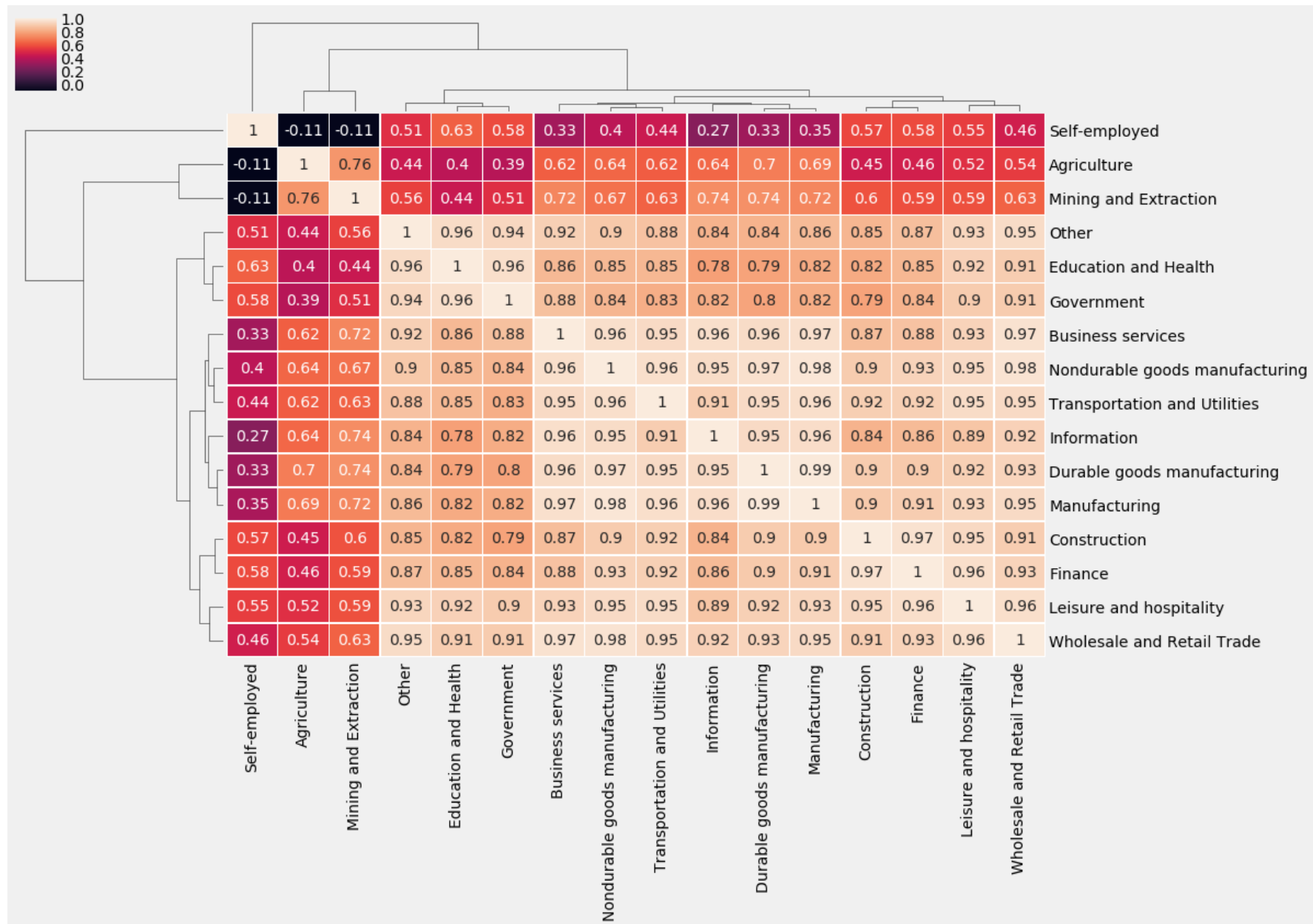
```
# Get correlation matrix of the seasonality_df DataFrame
trend_corr = trend_df.corr(method='spearman')

# 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!

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Congratulations!

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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!

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