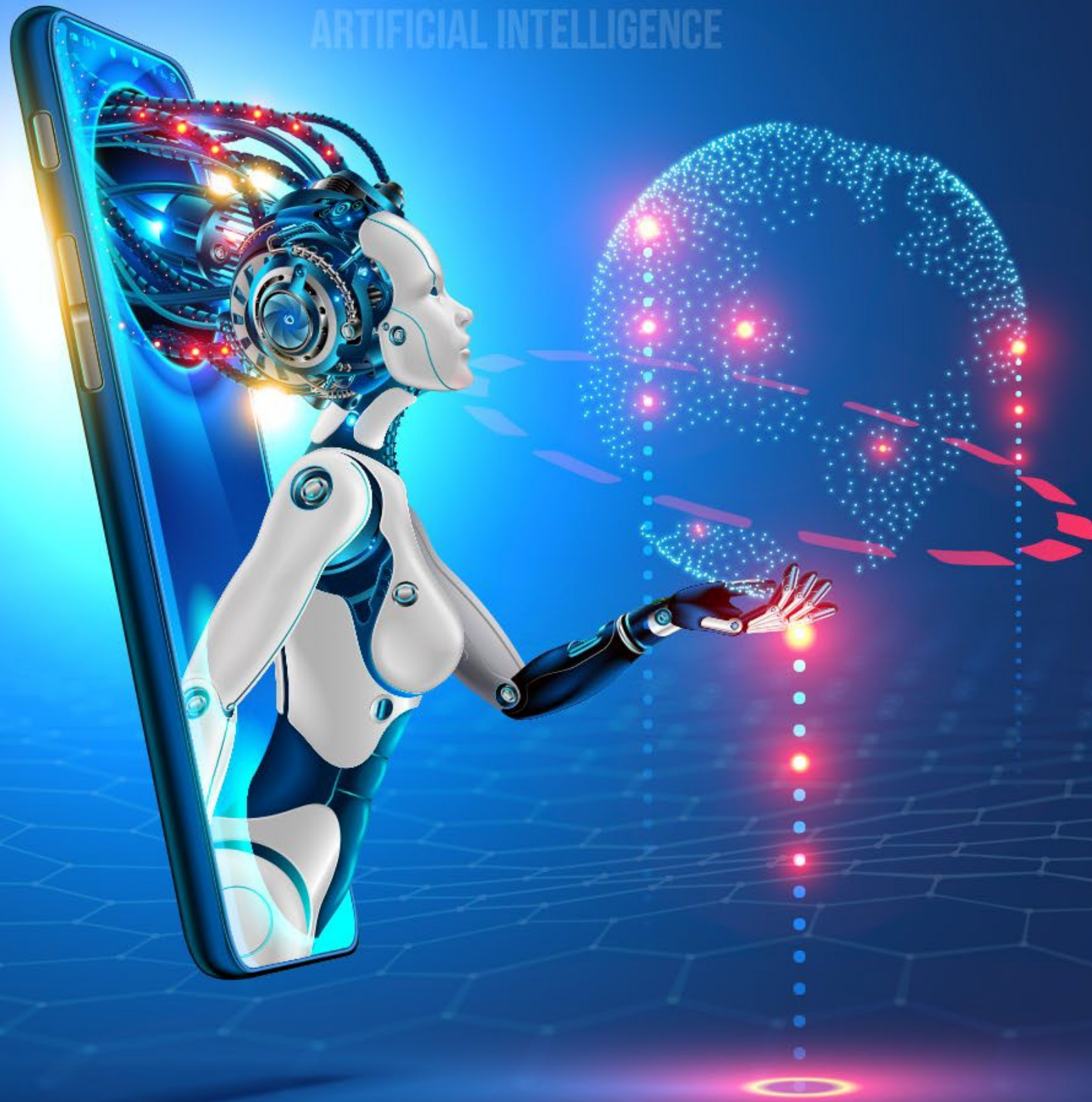


DATA AND ARTIFICIAL INTELLIGENCE



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Data Science With Python

DATA AND ARTIFICIAL INTELLIGENCE



Machine Learning with Scikit -Learn

Learning Objectives

By the end of this lesson, you will be able to:

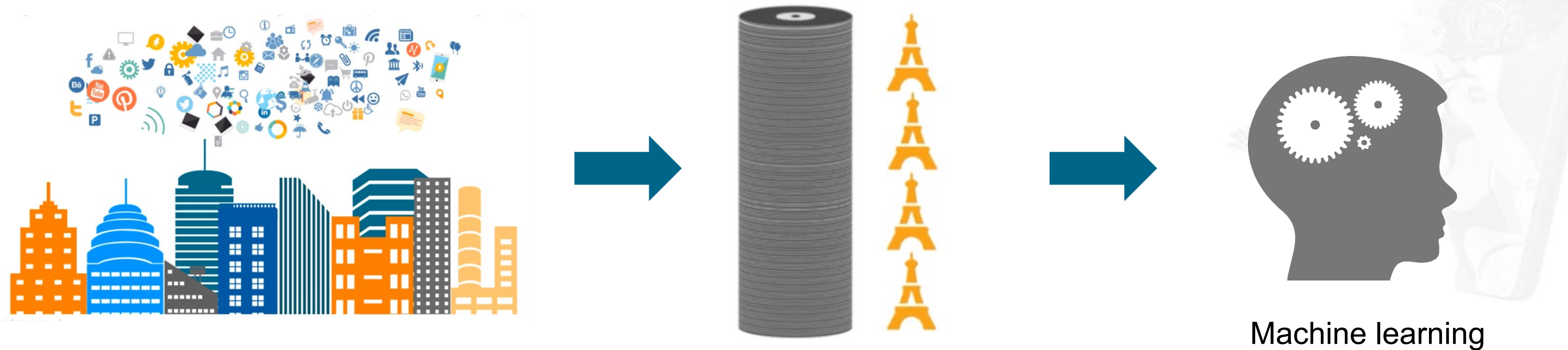
- 🕒 Define machine learning
- 🕒 Explain the machine learning approach
- 🕒 List relevant terminologies that help you understand a dataset
- 🕒 Discuss features of supervised and unsupervised learning models
- 🕒 Explain algorithms, such as regression, classification, clustering, and dimensionality reduction



Introduction to Machine Learning

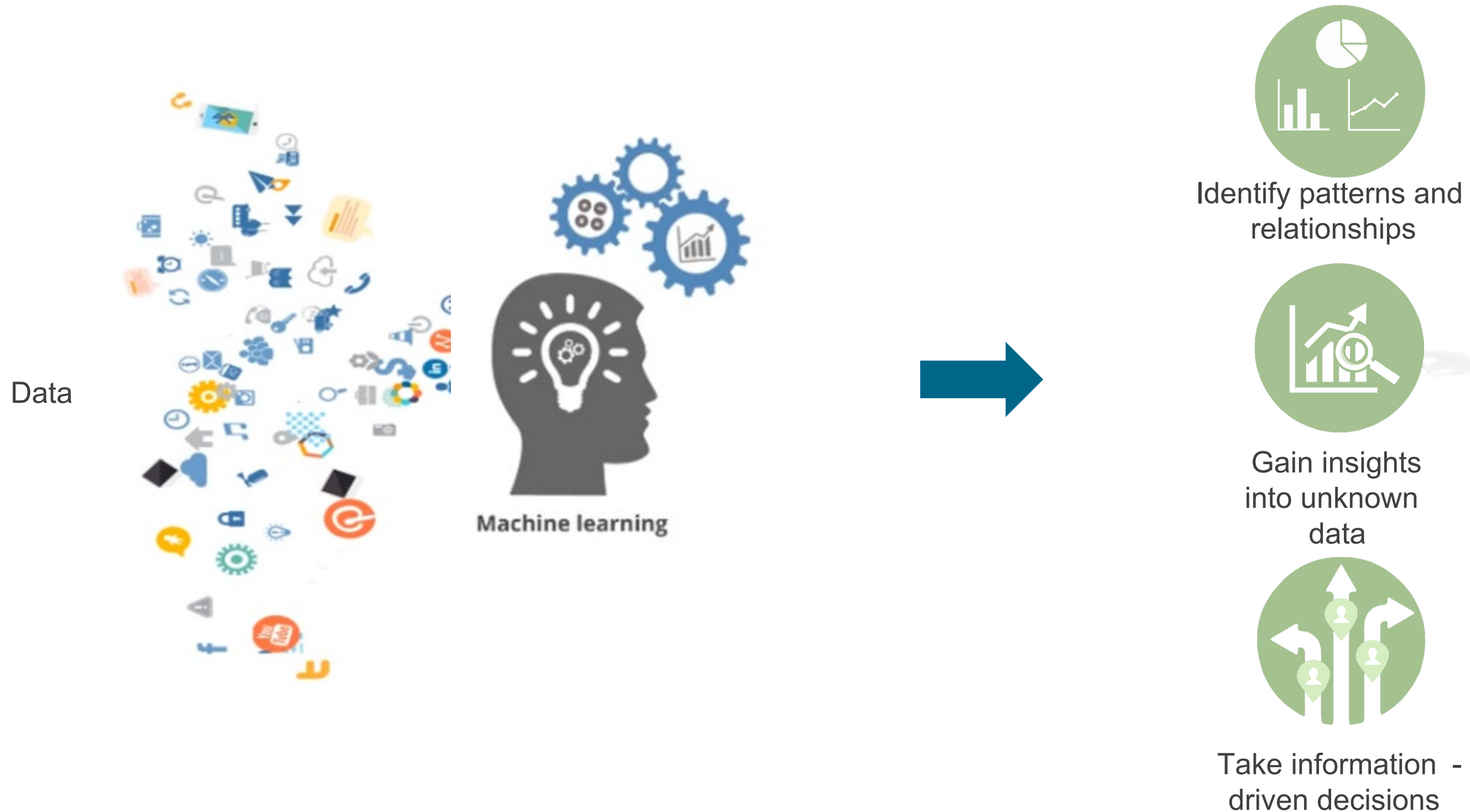
Why Machine Learning

If we stored the data generated in a day on Blu-ray disks and stacked them up, it would be equal to the height of four Eiffel towers. Machine learning helps analyze this data easily and quickly.



Purpose of Machine Learning

Machine learning is a great tool to analyze data, find hidden data patterns and relationships, and extract information to enable information -driven decisions and provide insights.



Machine Learning Terminology

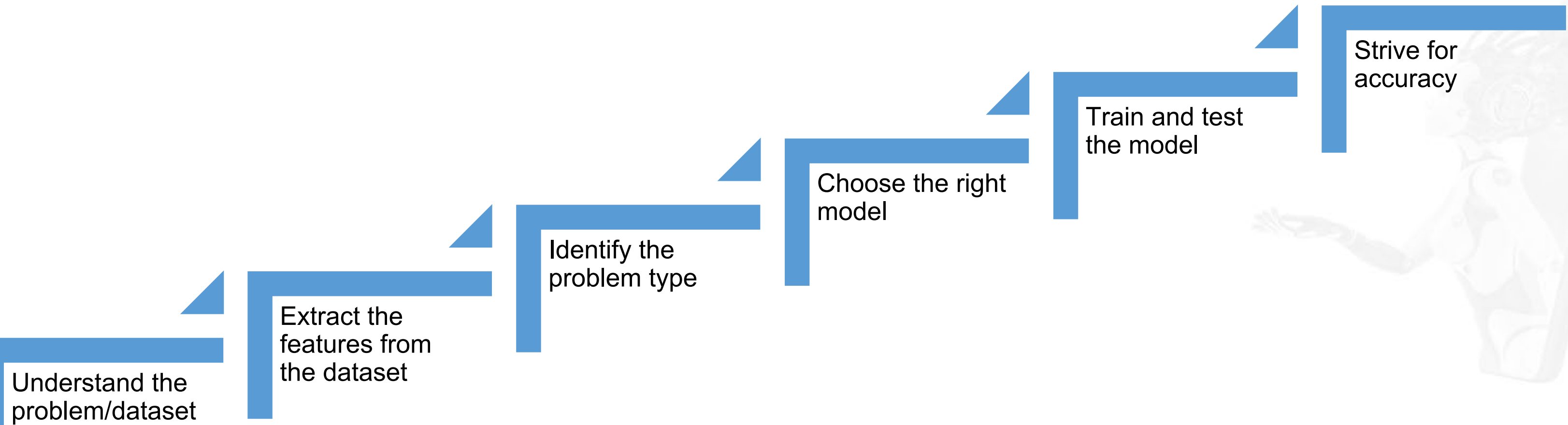
These are some machine learning terminologies that you will come across in this lesson:



Machine Learning Approach

Machine Learning Approach

The machine learning approach starts with either a problem that you need to solve or a given dataset that you need to analyze.



Steps 1 and 2: Understand the Dataset and Extract Its Features

Let us look at a dataset and understand its features in terms of machine learning.

Features (attributes)	Education (Yrs.)	Professional Training (Yes/No)	Hourly Rate (USD)	Response (label)
Observations (records)	16	1	90	
	15	0	65	
	12	1	70	
	18	1	130	
	16	0	110	
	16	1	100	
	15	1	105	
	31	0	70	
Predictors				

Steps 3 and 4: Identify the Problem Type and Learning Model

Machine learning can either be supervised or unsupervised. The problem type should be selected based on the type of learning model.

Concept

Problem Types

Example

Supervised Learning

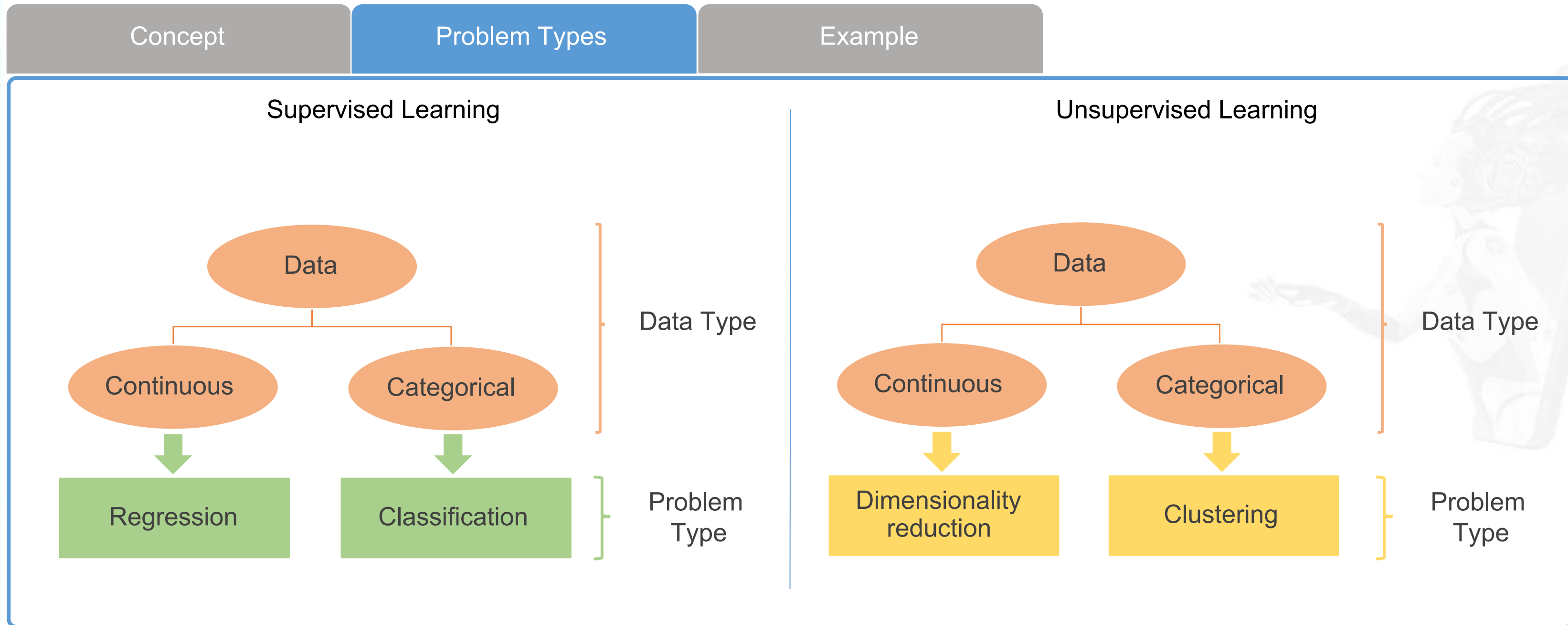
- In supervised learning, the dataset used to train a model should have observations, features, and responses. The model is trained to predict the “right” response for a given set of data points.
- Supervised learning models are used to predict an outcome.
- The goal of this model is to “generalize” a dataset so that the “general rule” can be applied to new data as well.

Unsupervised Learning

- In unsupervised learning, the response or the outcome of the data is not known.
- Unsupervised learning models are used to identify and visualize patterns in data by grouping similar types of data.
- The goal of this model is to “represent” data in a way that meaningful information can be extracted.

Steps 3 and 4: Identify the Problem Type and Learning Model

Data can either be continuous or categorical. Based on whether it is supervised or unsupervised learning, the problem type will differ.



Steps 3 and 4: Identify the Problem Type and Learning Model

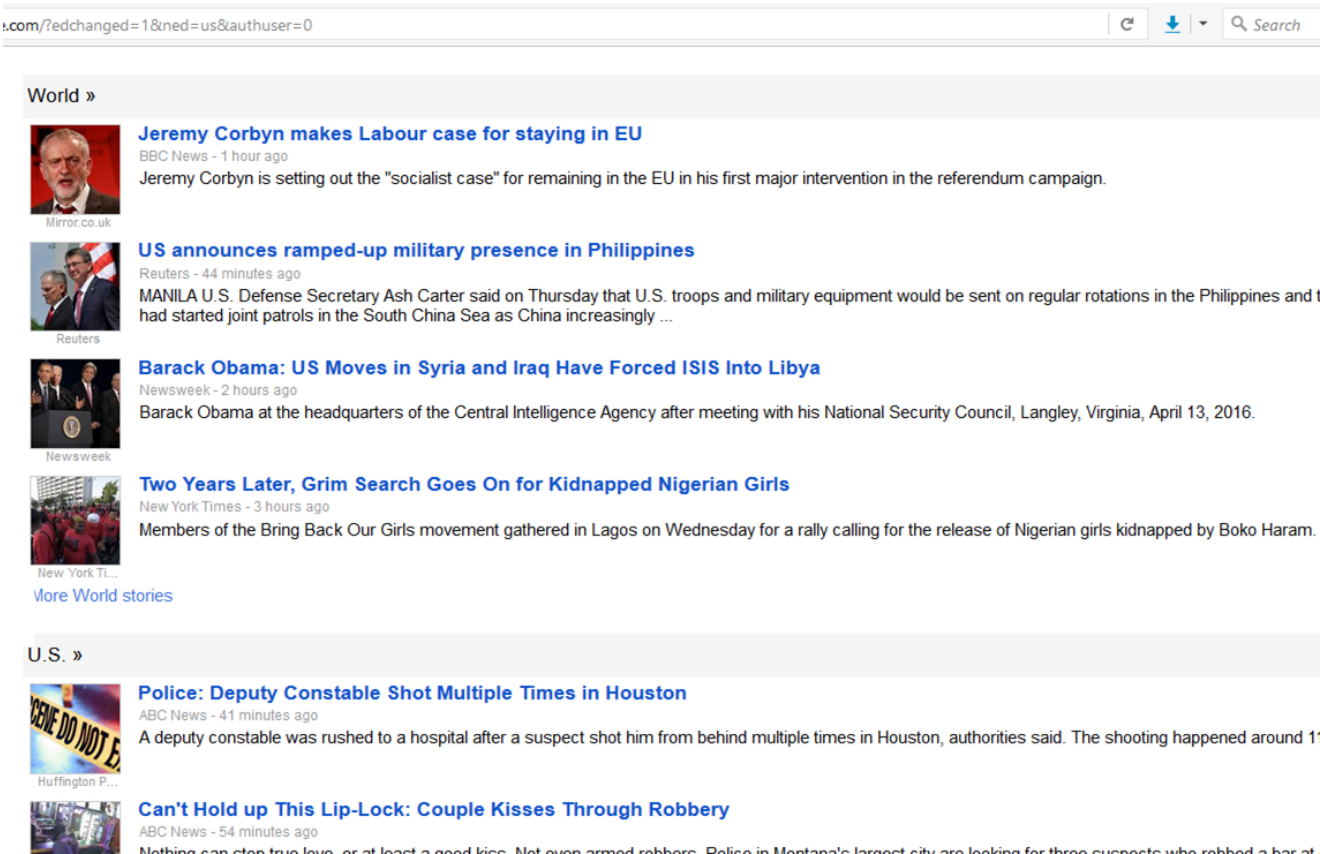
Some examples of supervised and unsupervised learning models are:

Concept

Problem Types

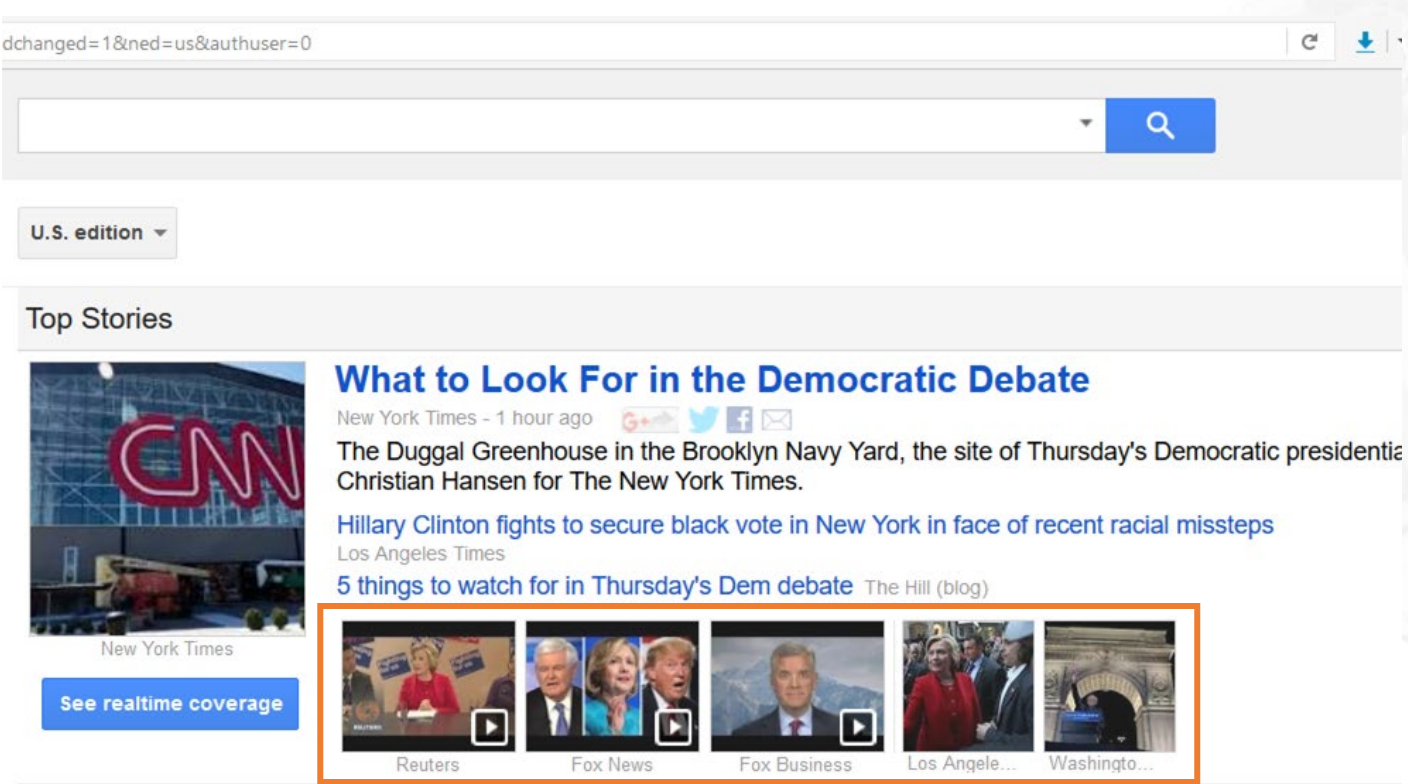
Example

Supervised Learning



Categories of news based on the topics

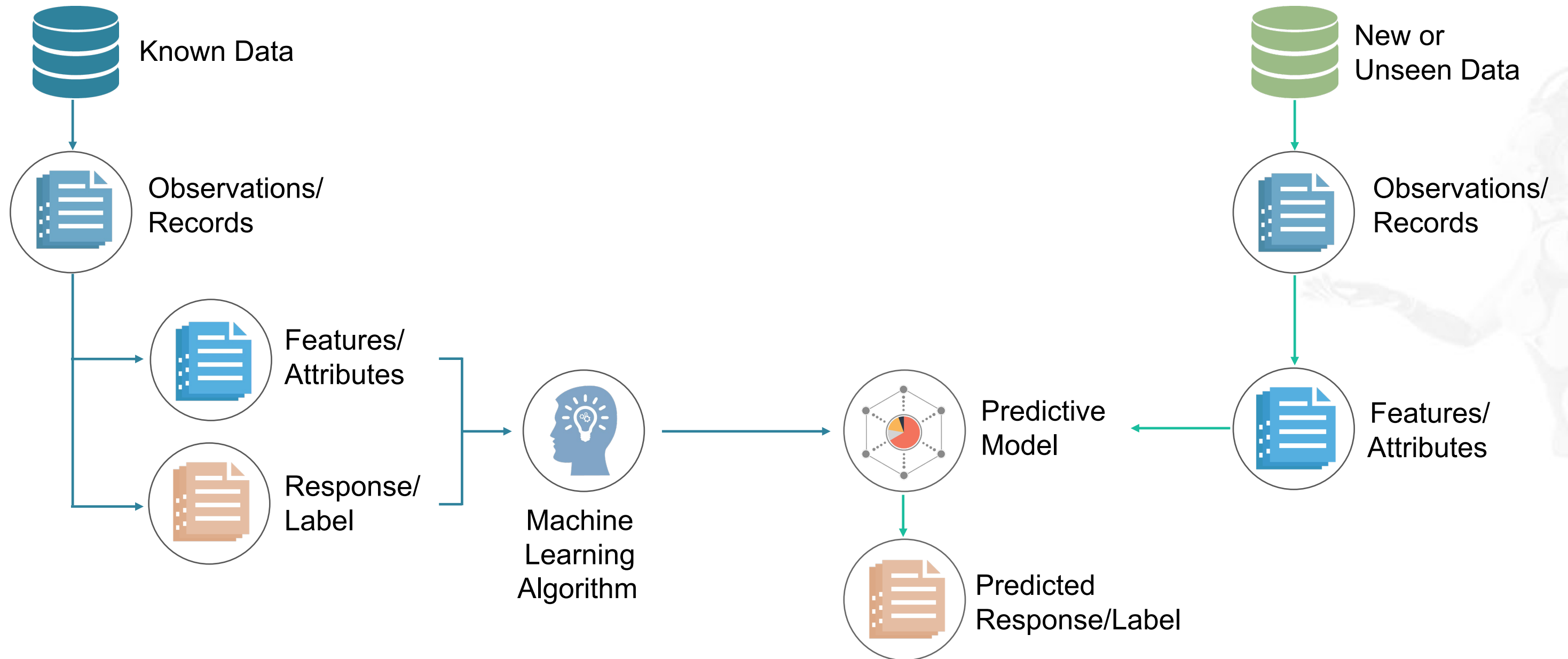
Unsupervised Learning



Grouping of similar stories on different news networks

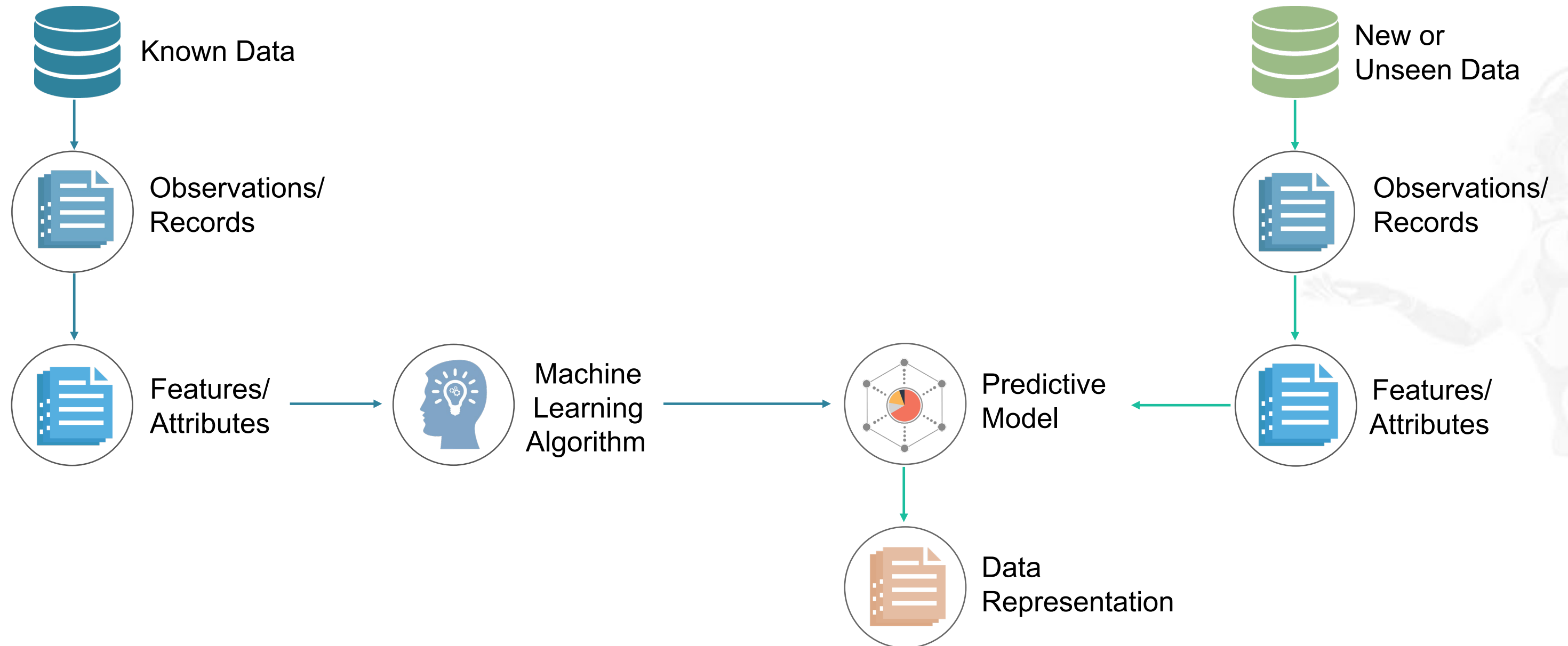
Working of Supervised Learning Model

In supervised learning, a known dataset with observations, features, and response is used to create and train a machine learning algorithm. A predictive model, built on top of this algorithm, is then used to predict the response for a new dataset that has the same features.



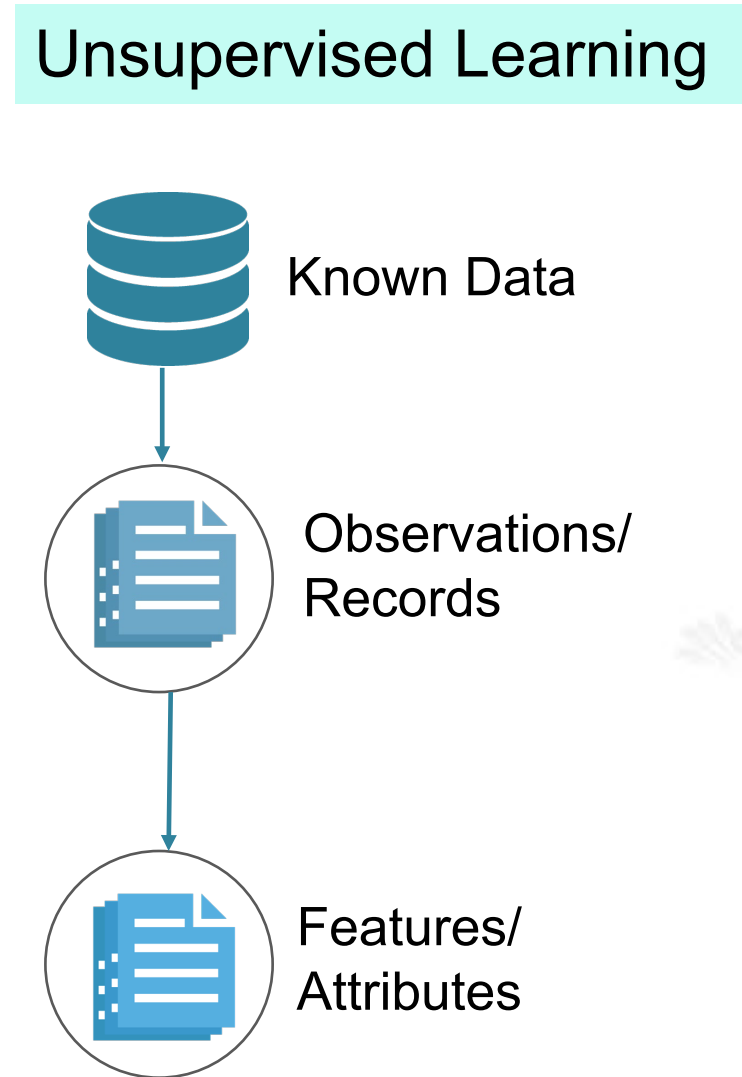
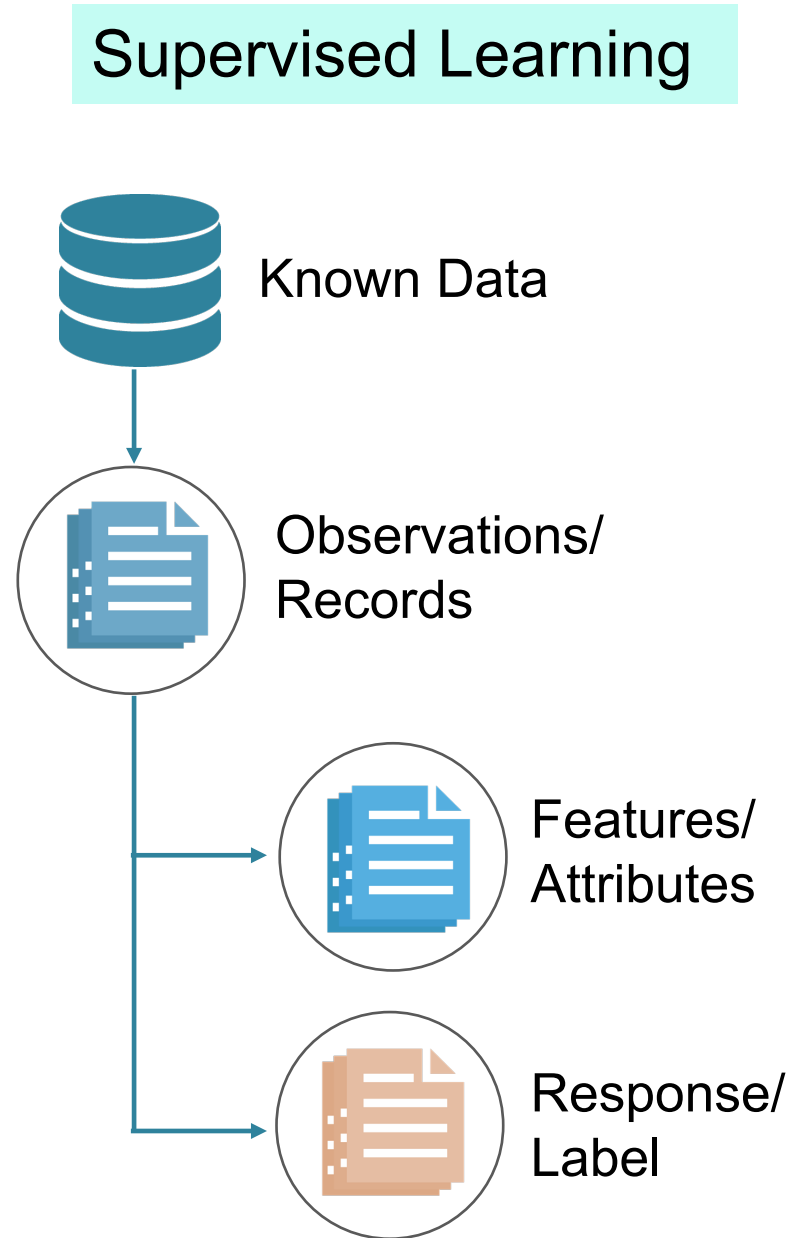
Working of Unsupervised Learning Model

In unsupervised learning, a known dataset has a set of observations with features. But the response is not known. The predictive model uses these features to identify how to classify and represent the data points of new or unseen data.

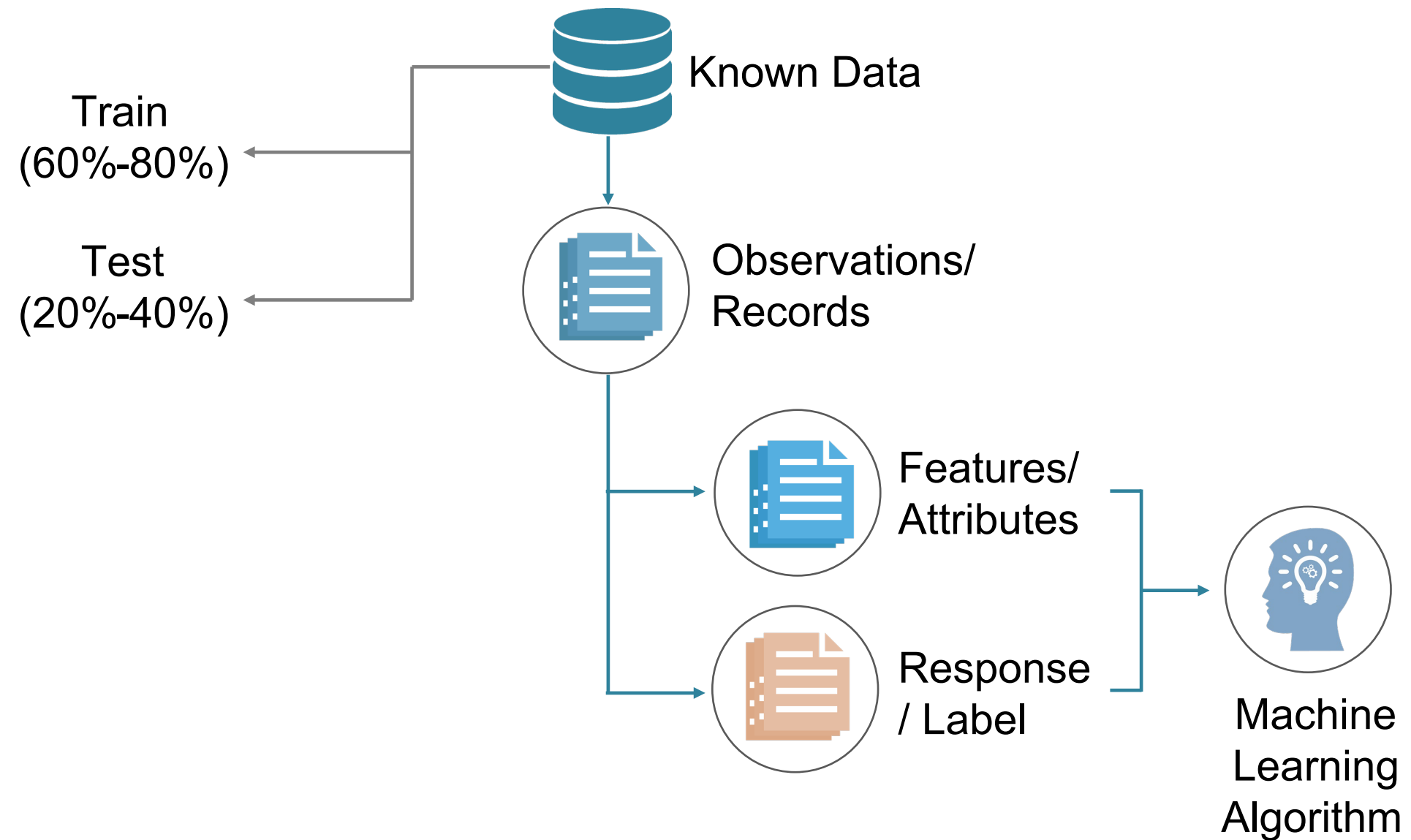


Steps 5 and 6: Train, Test, and Optimize the Model

To train supervised learning models, data analysts usually divide a known dataset into training and testing sets.

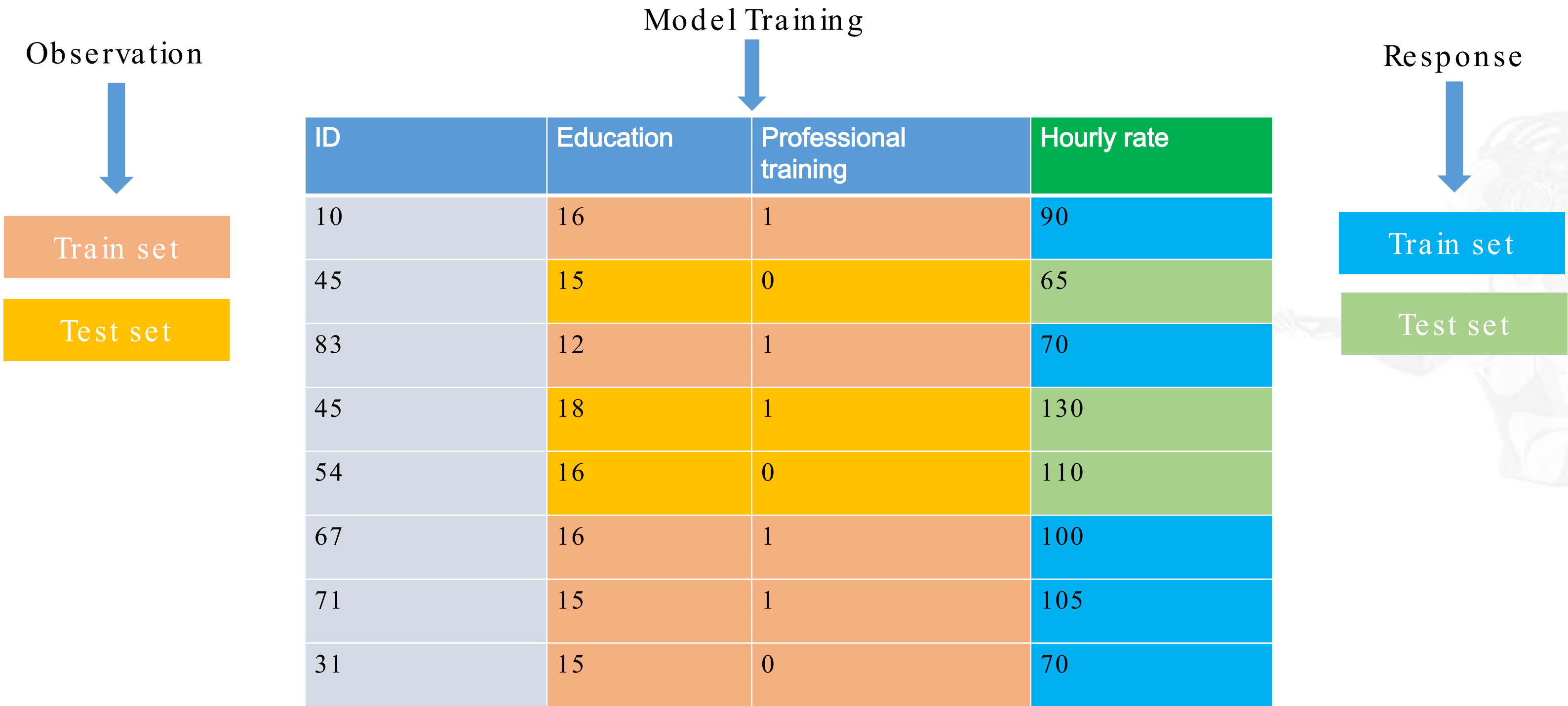


Steps 5 and 6: Train, Test, and Optimize the Model



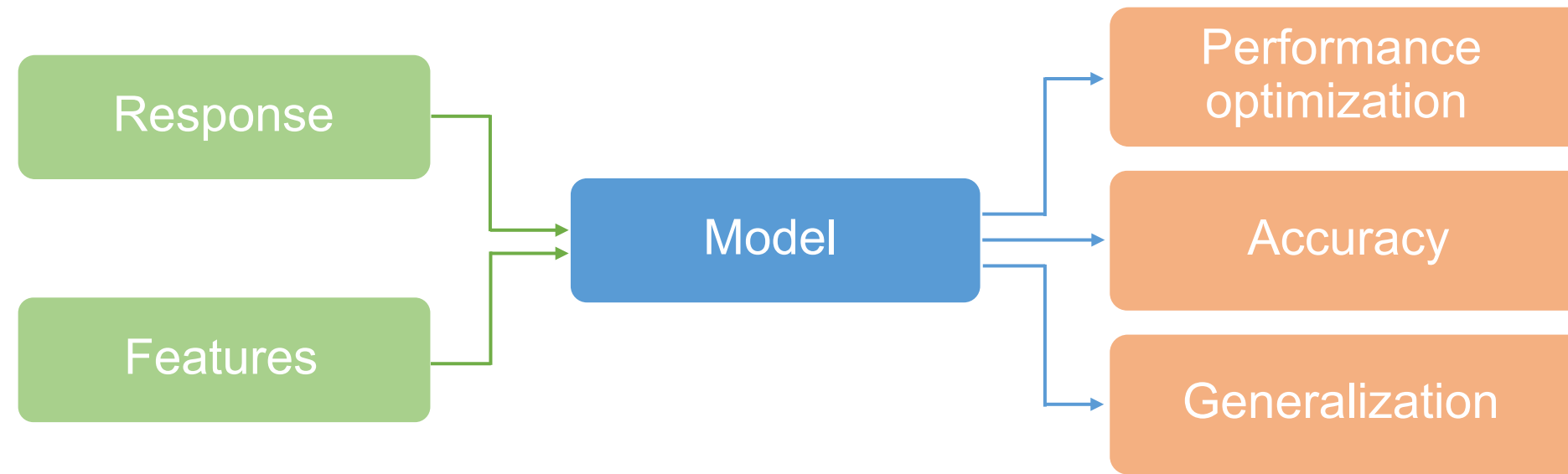
Steps 5 and 6: Train, Test, and Optimize the Model

Let us look at an example to see how the split approach works.



Supervised Learning Model Considerations

Some considerations of supervised and unsupervised learning models are shown here.

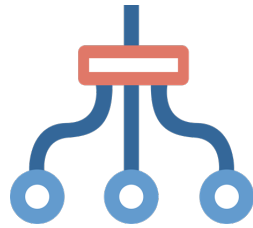


DATA AND ARTIFICIAL INTELLIGENCE

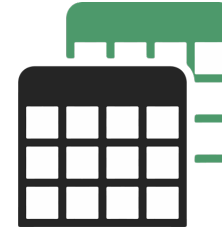
Scikit -Learn

Scikit -Learn

Scikit is a powerful and modern machine learning Python library for fully and semi - automated data analysis and information extraction.



Efficient tools to identify and organize problems (Supervised/ Unsupervised)



Free and open datasets



Rich set of libraries for learning and predicting



Model support for every problem type



Model persistence

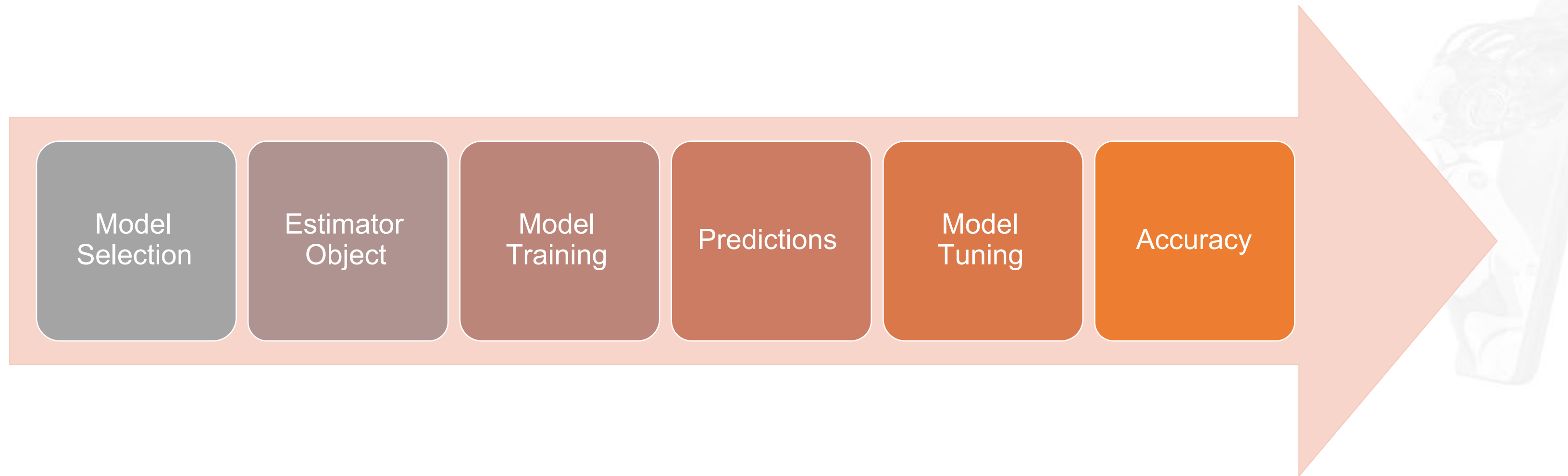


open source
initiative

Open source community and vendor support

Scikit -Learn: Problem -Solution Approach

Scikit-learn helps Data Scientists organize their work through its problem -solution approach.



Scikit -Learn: Problem -Solution Considerations

While working with a Scikit -Learn dataset or loading your own data to Scikit -Learn, always consider these points:



Create separate objects for feature and response



Ensure that features and response have only numeric values



Features and response should be in the form of a NumPy ndarray



Since features and response would be in the form of arrays, they would have shapes and sizes

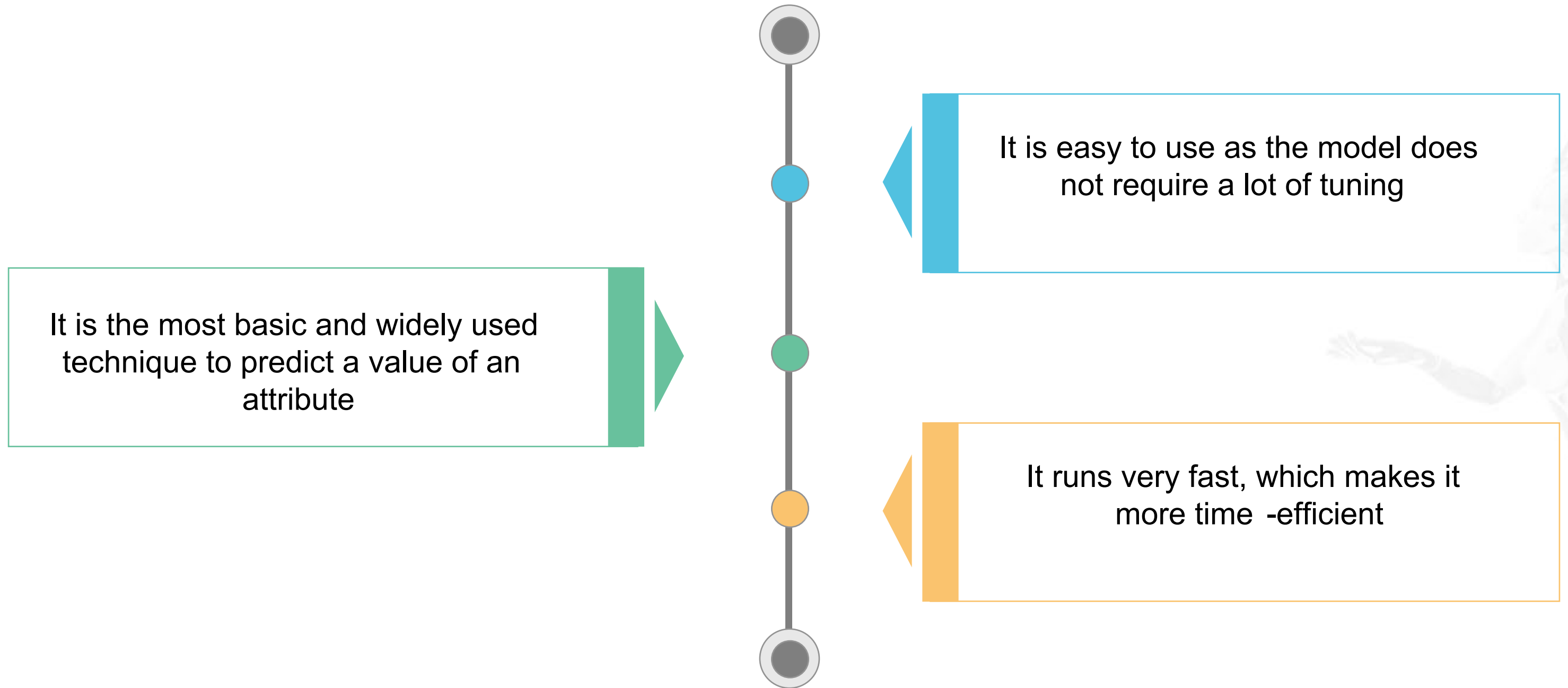


Features are always mapped as x , and response is mapped as y

Supervised Learning Models: Linear Regression

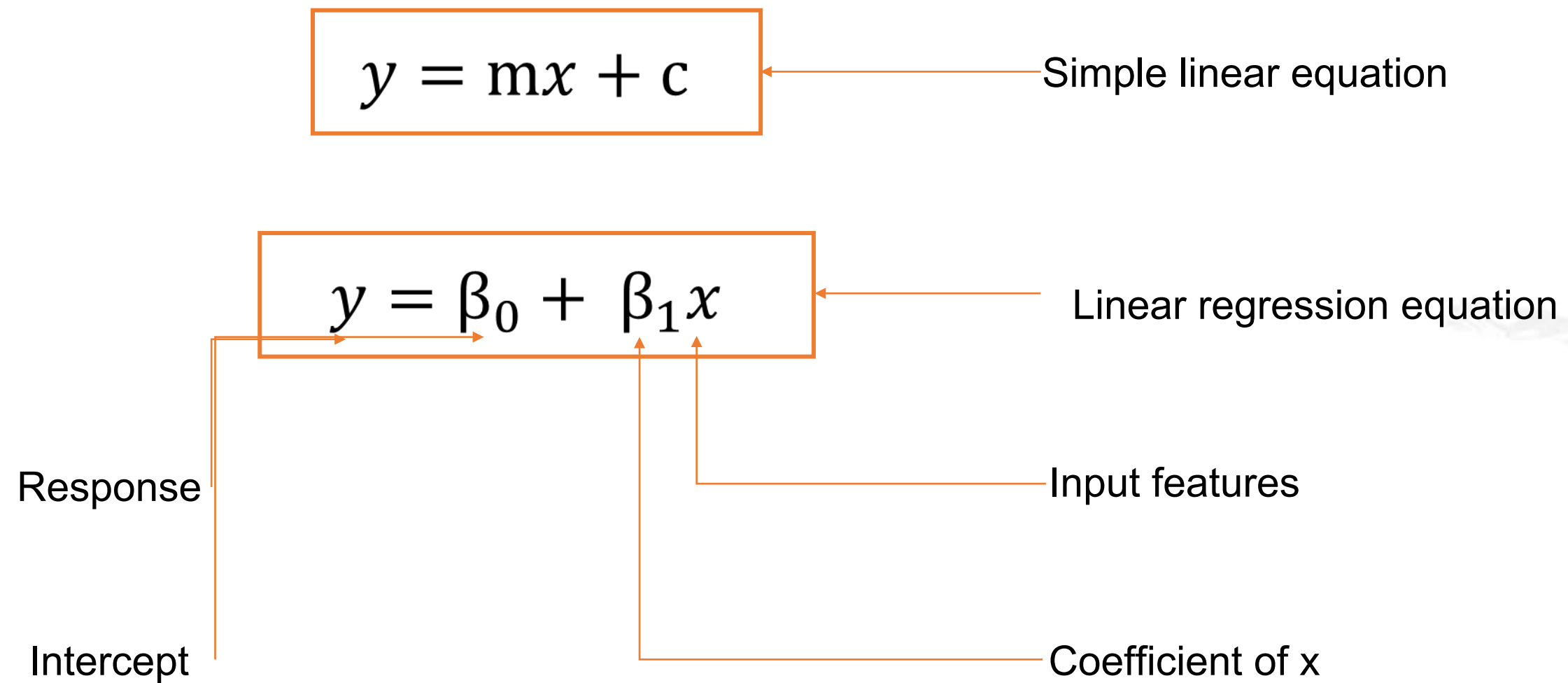
Supervised Learning Models: Linear Regression

Linear regression is a supervised learning model used to analyze continuous data.



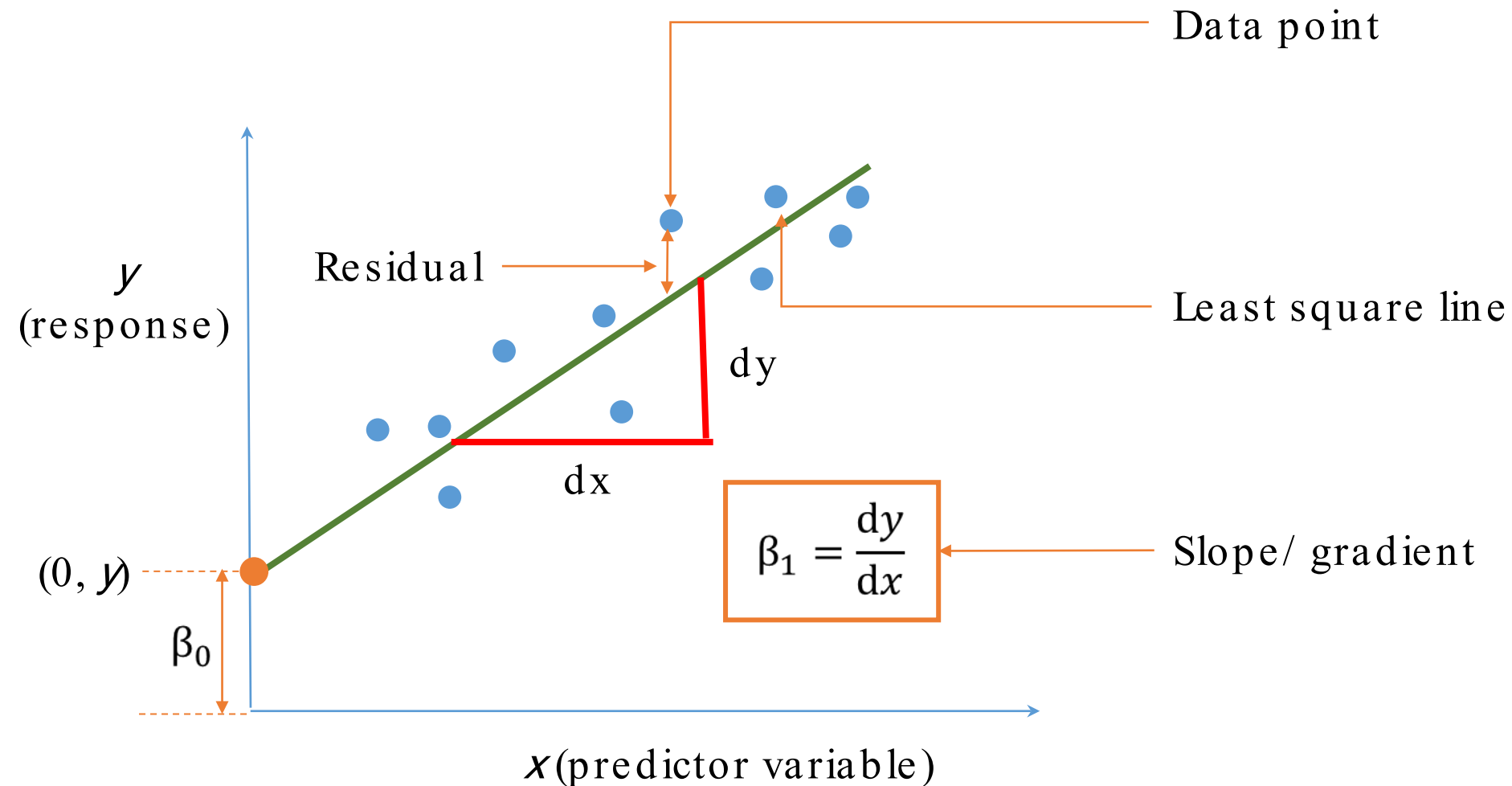
Supervised Learning Models: Linear Regression

The linear regression equation is based on the formula for a simple linear equation.



Supervised Learning Models: Linear Regression

Linear regression is the most basic technique to predict a value of an attribute.



$$y = \beta_0 + \beta_1 x + u$$

Labels for the equation components:

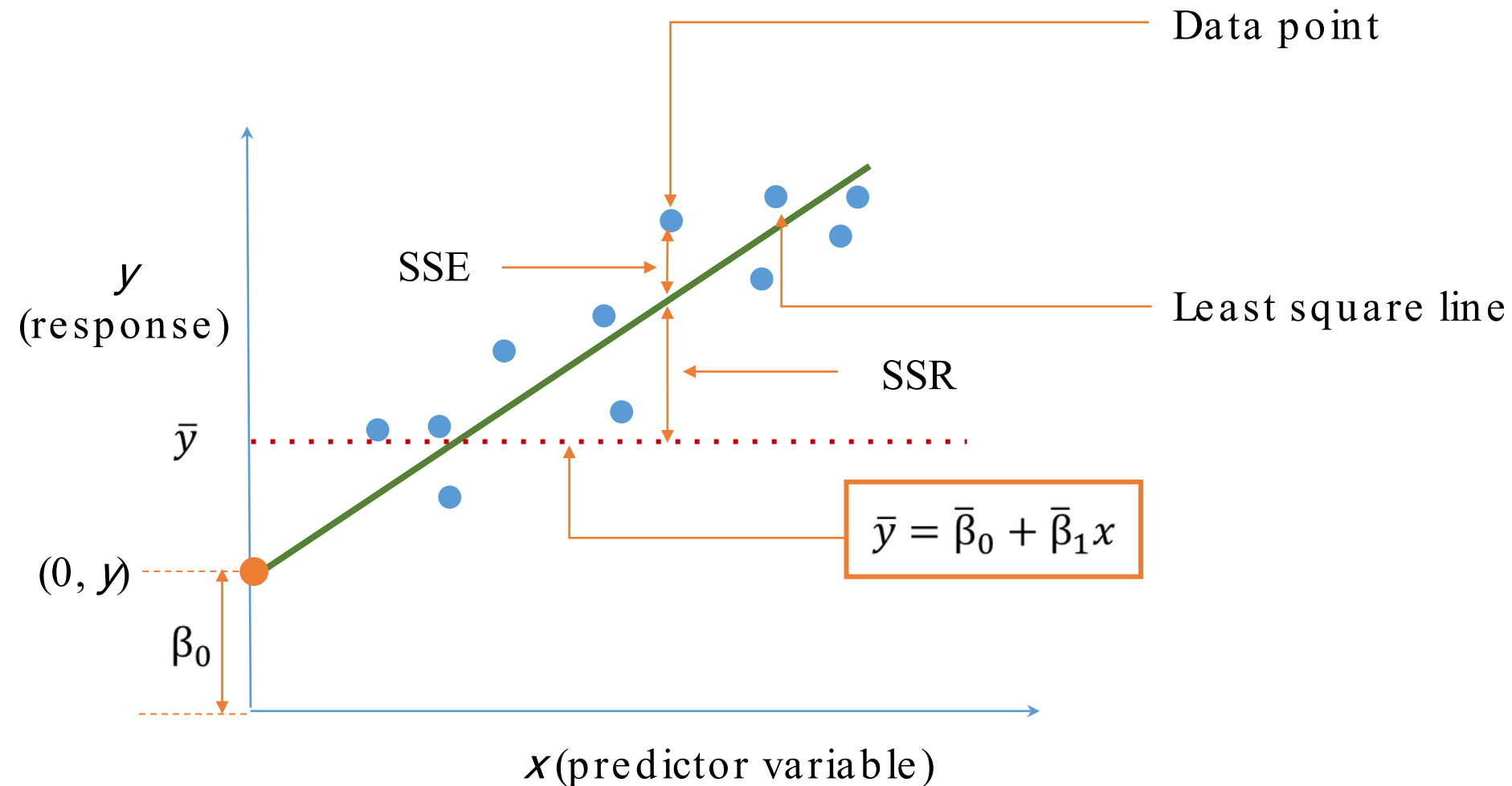
- y : Actual value
- $\beta_0 + \beta_1 x$: Predicted value
- u : Residual



The attributes are usually fitted using the “least square” approach.

Supervised Learning Models: Linear Regression

Smaller the value of SSR or SSE, the more accurate the prediction will be, which would make the model the best fit.



$$y = \beta_0 + \beta_1 x + u$$

$$SSR = \sum (\hat{y}_i - \bar{y})^2$$

Regression of sum of squares

$$SSE = \sum (y_i - \hat{y}_i)^2$$

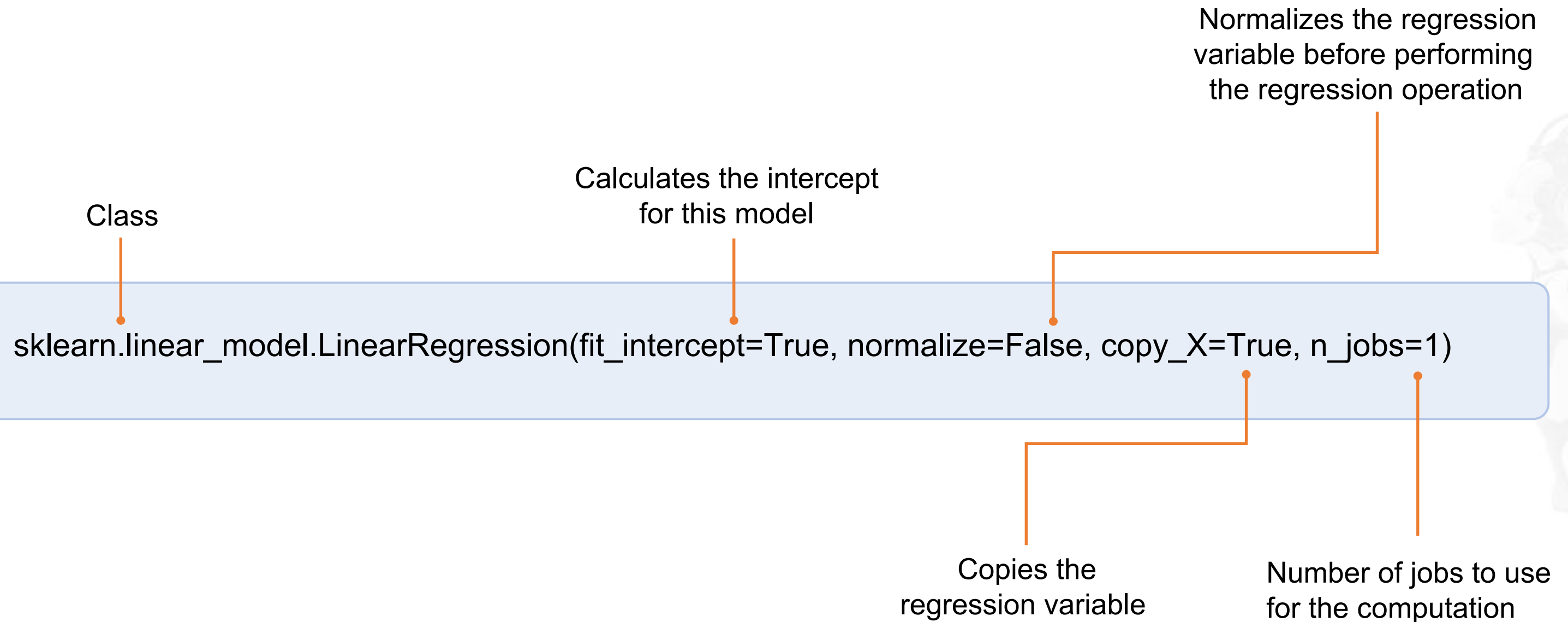
Error of sum of squares



The attributes are usually fitted using the “least square” approach.

Supervised Learning Models: Linear Regression

Let us see how linear regression works in Scikit -Learn.



Loading a Dataset



Problem Statement: Demonstrate how to load a built-in SciKit-learn dataset.

Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login**.

ASSISTED PRACTICE

Linear Regression Model



Problem Statement: Demonstrate how to create and train a linear regression model

Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login**.

ASSISTED PRACTICE

Supervised Learning Models: Logistic Regression

Supervised Learning Models: Logistic Regression

Logistic regression is a generalization of the linear regression model used for classification problems.

$$\pi = \Pr(y = 1|x) = \frac{e^{\beta_0 + \beta_1 x}}{1 + e^{\beta_0 + \beta_1 x}}$$

Probability of $y = 1$, given x

Change in the log-odds for a unit change in x



The purpose of K-NN is to predict the class for each observation.

Supervised Learning Models: Logistic Regression

Logistic regression is a generalization of the linear regression model used for classification problems.

$$\text{Odds} = \frac{\pi}{1 - \pi}$$

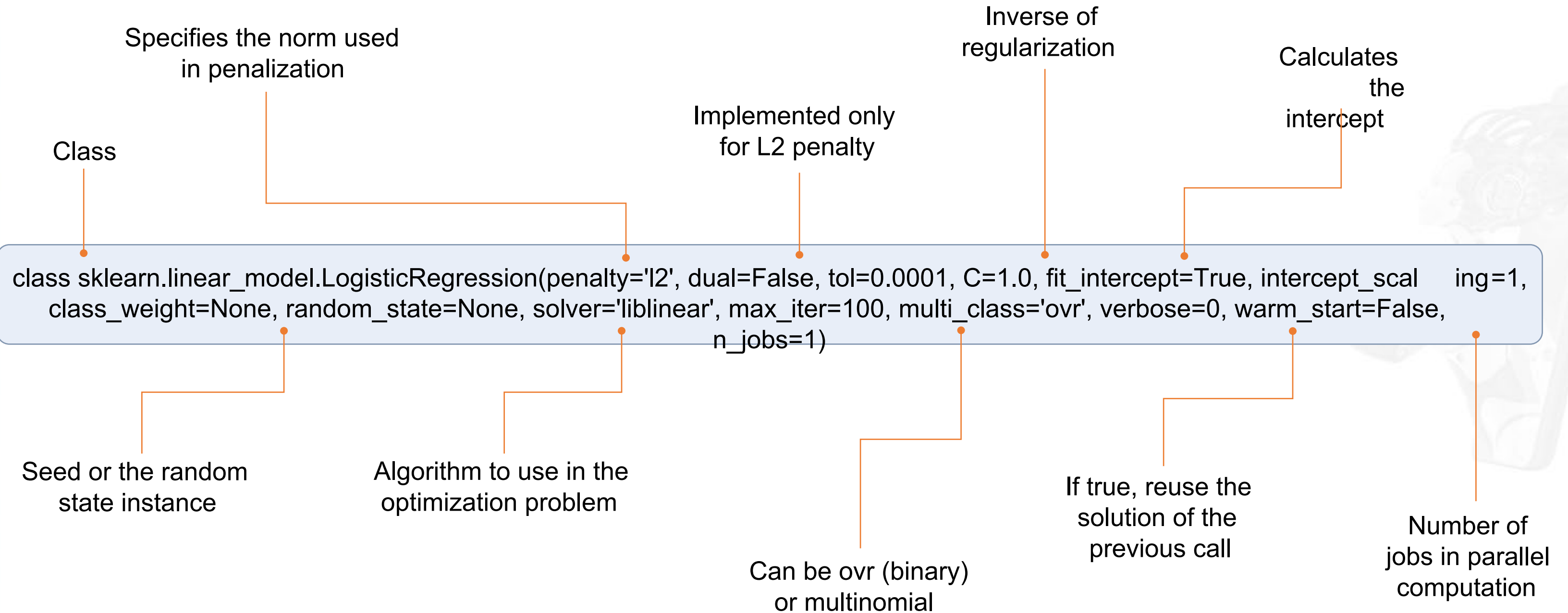
Probability 

$$\log\left(\frac{\pi}{1 - \pi}\right) = \log\left(e^{\beta_0 + \beta_1 x}\right) = \beta_0 + \beta_1 x$$

Logarithm of odds 

 Linear regression

Supervised Learning Models: Logistic Regression



Supervised Learning Models: : K -Nearest Neighbors

Supervised Learning Models: K -Nearest Neighbors (K -NN)

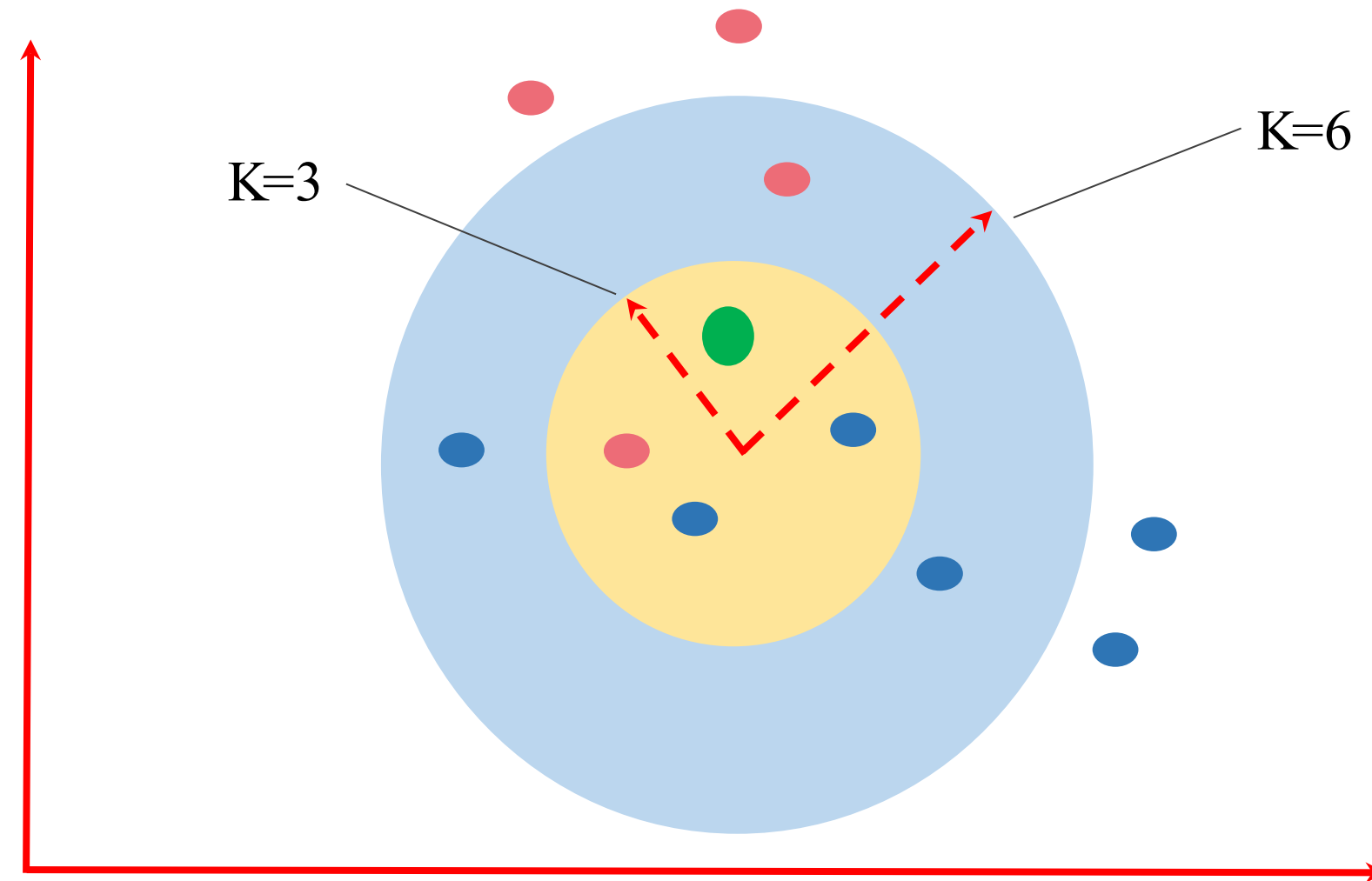
K-Nearest Neighbors, or K -NN, is one of the simplest machine learning algorithms used for both classification and regression problem types.

Features
(Attributes)



Education (Yrs.)	Professional Training (Yes/No)	Hourly Rate (USD)
16	1	90
15	0	65
12	1	70
18	1	130
16	0	110
16	1	100
15	1	105
31	0	70

Supervised Learning Models: K -Nearest Neighbors



If you are using this method for binary classification, choose an odd number for k to avoid the case of a

tied

Supervised Learning Models: K -Nearest Neighbors

K-means finds the best centroids by alternatively assigning random centroids to a dataset and selecting mean data points from the resulting clusters to form new centroids. It continues this process iteratively until the model is optimized.

Features
(Attributes)

Response
(label)

Education (Yrs.)	Professional Training (Yes/No)	Hourly Rate (USD)
16	1	90
15	0	65
12	1	70
18	1	130
16	0	110
16	1	100
15	1	105
31	0	70

K-NN and Logistic Regression Models



Problem Statement: Demonstrate the use of K -NN and logistic regression models

Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login** .

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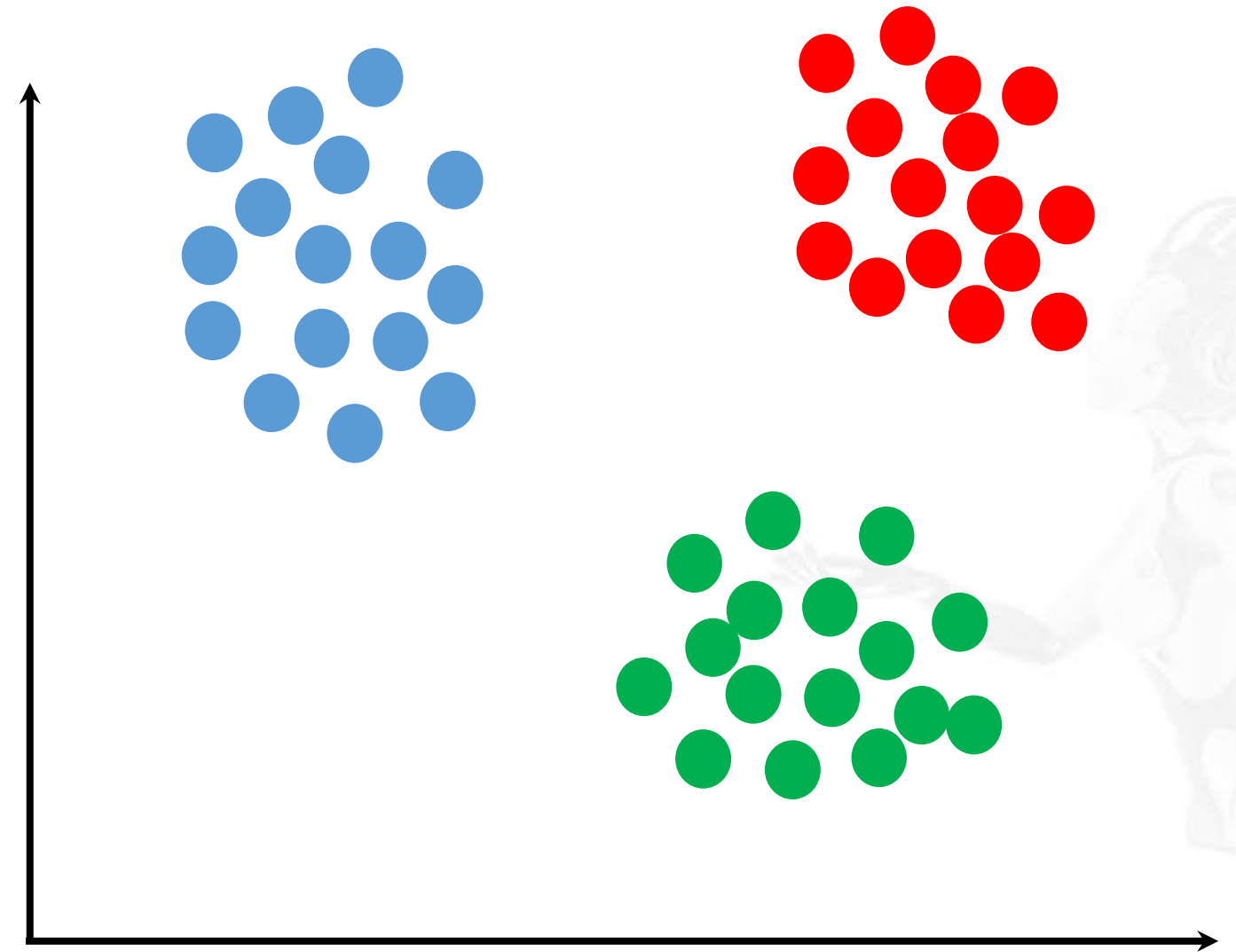
Unsupervised Learning Models: Clustering

Unsupervised Learning Models: Clustering

A cluster is a group of similar data points.

Clustering is used to:

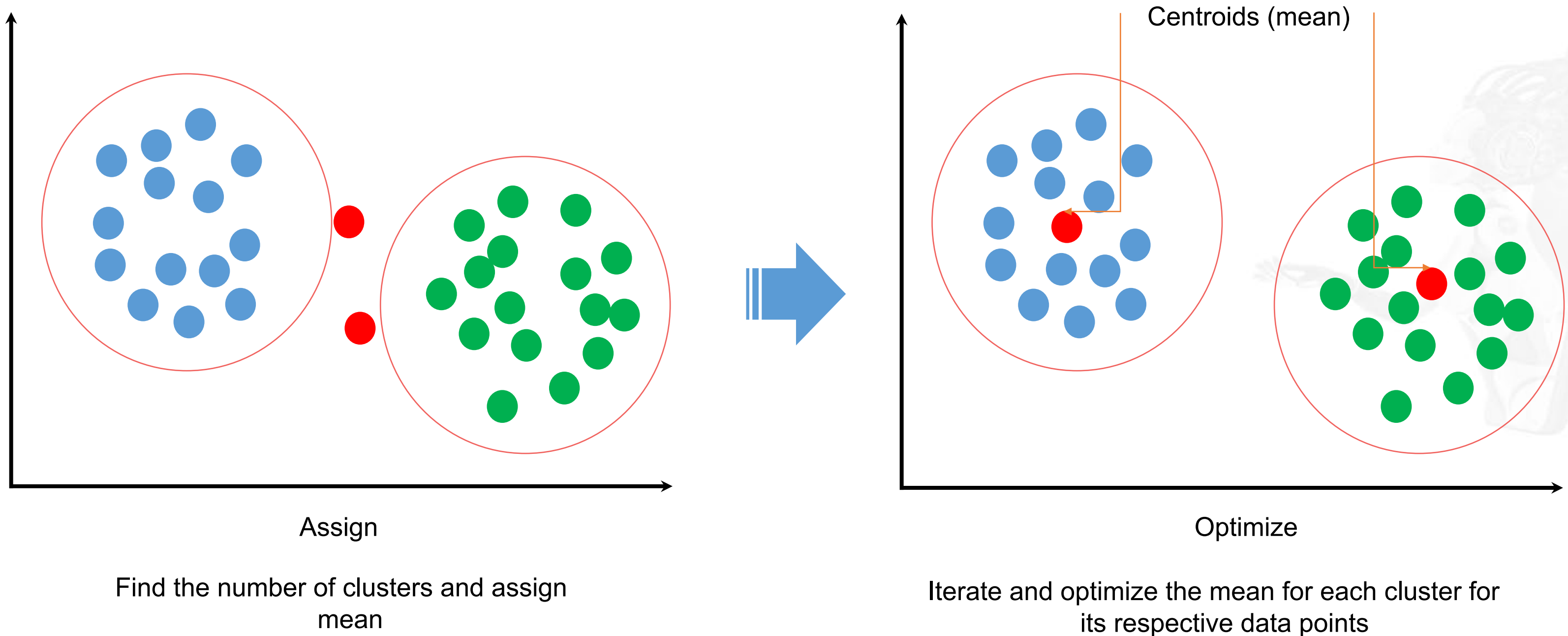
- Extract the structure of the data
- Identify groups in the data



Greater similarity between data points results in better clustering.

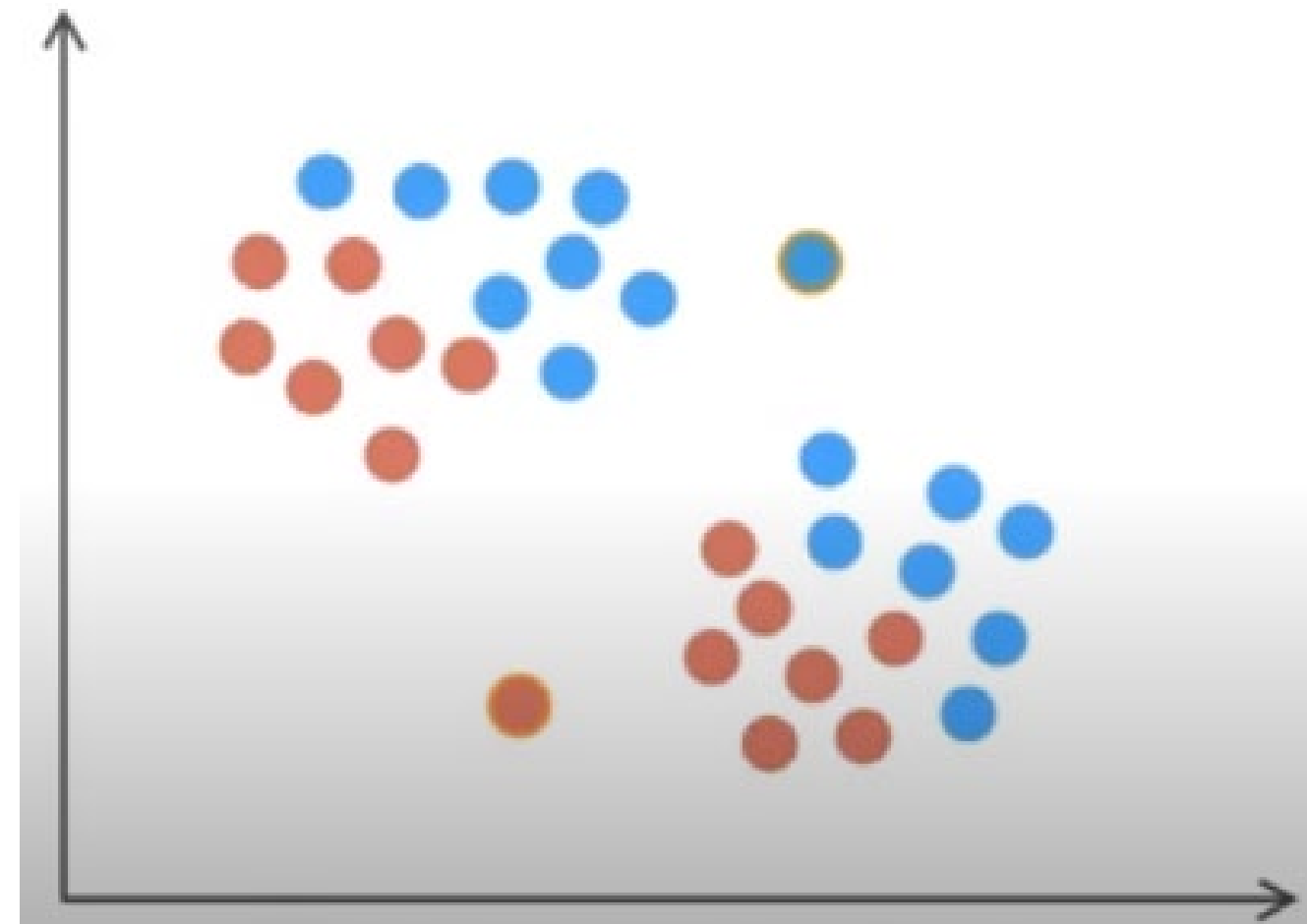
Unsupervised Learning Models: K -Means Clustering

K-means finds the best centroids by alternatively assigning random centroids to a dataset and selecting mean data points from the resulting clusters to form new centroids. It continues this process iteratively until the model is optimized.



Unsupervised Learning Models: K -Means Clustering

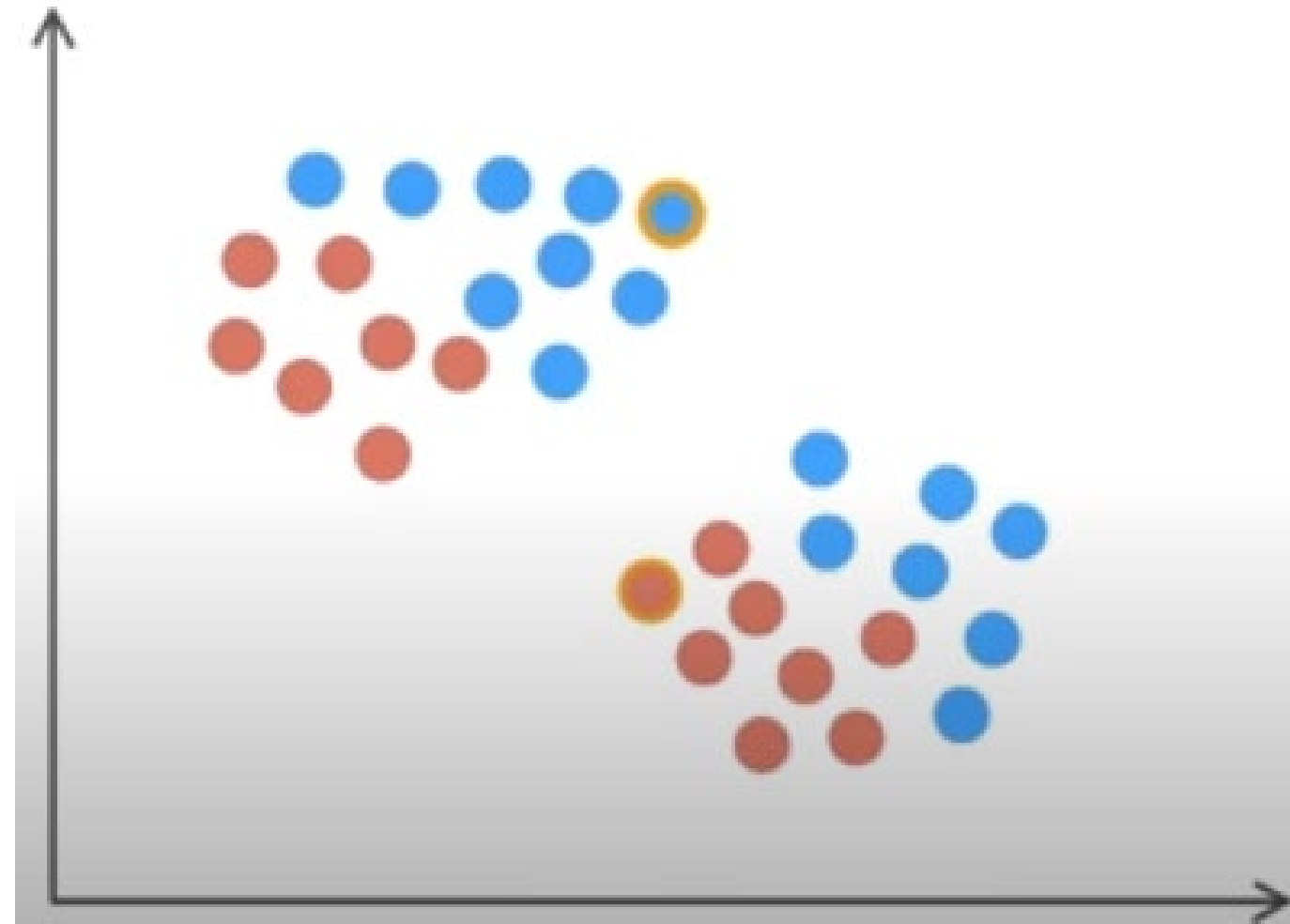
K-means finds the best centroids by alternatively assigning random centroids to a dataset and selecting mean data points from the resulting clusters to form new centroids. It continues this process iteratively until the model is optimized.



Assign data points to the centroids

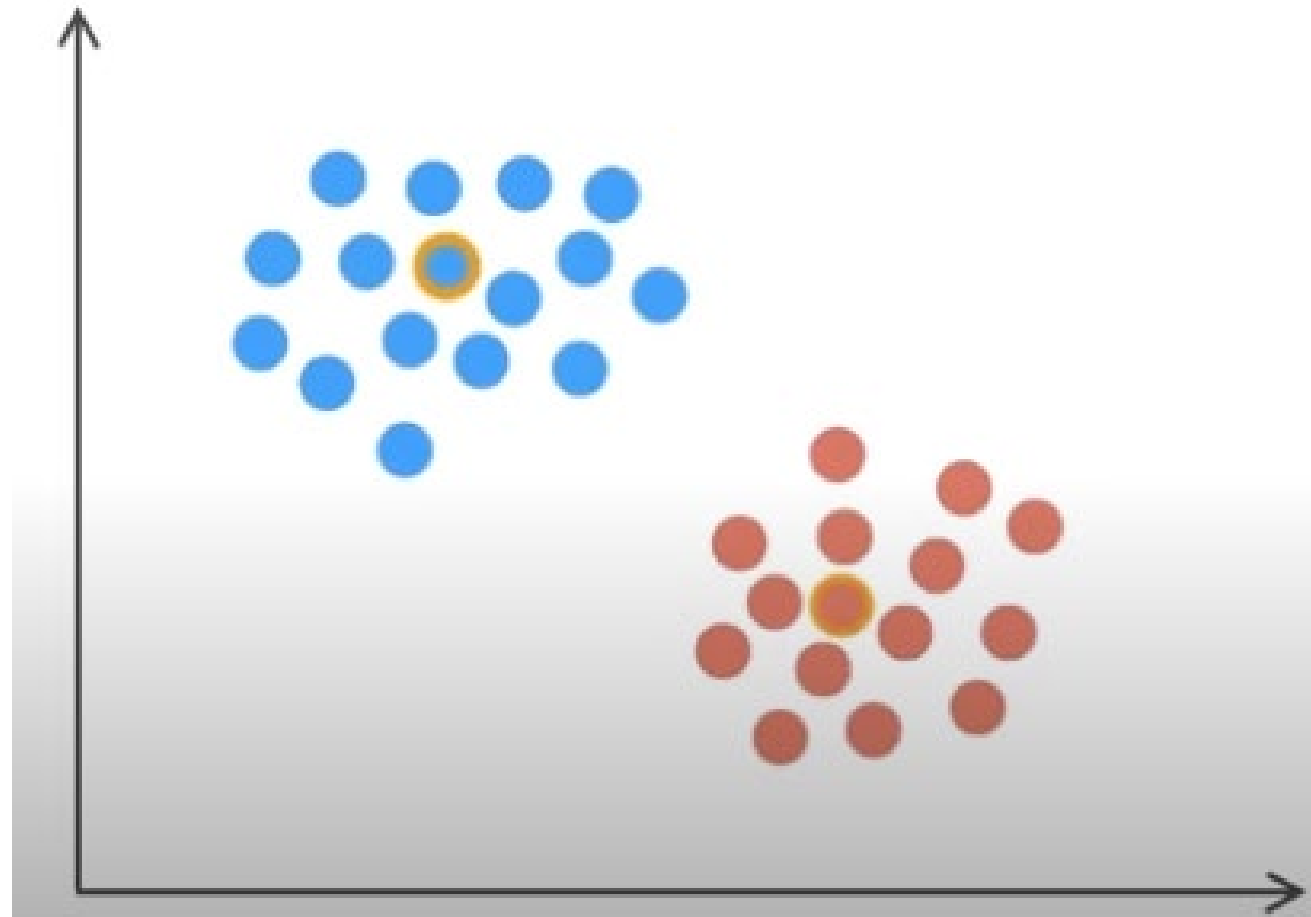
Unsupervised Learning Models: K -Means Clustering

Choose a mean from each cluster as a centroid



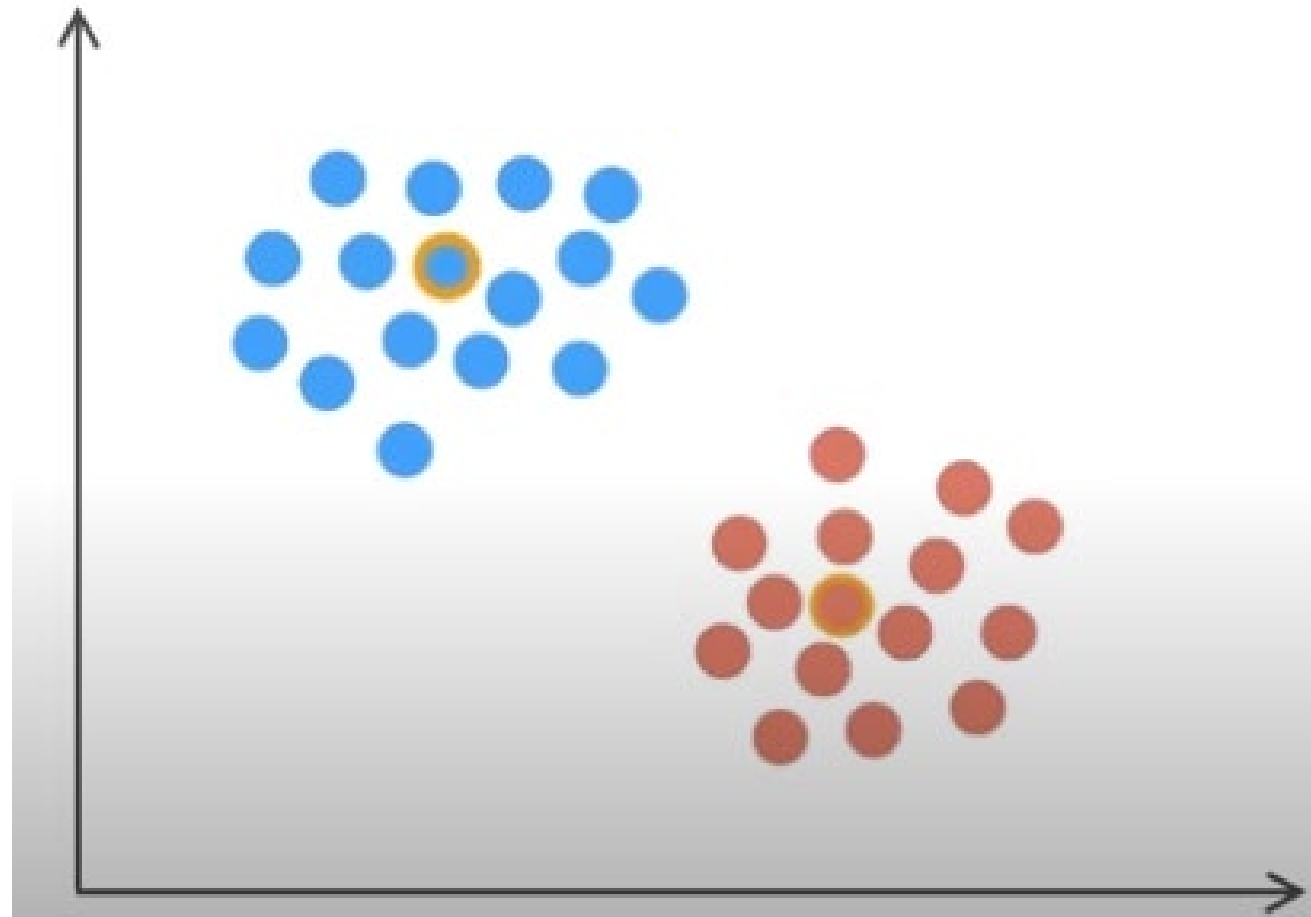
Unsupervised Learning Models: K -Means Clustering

Reassign data points to new centroids



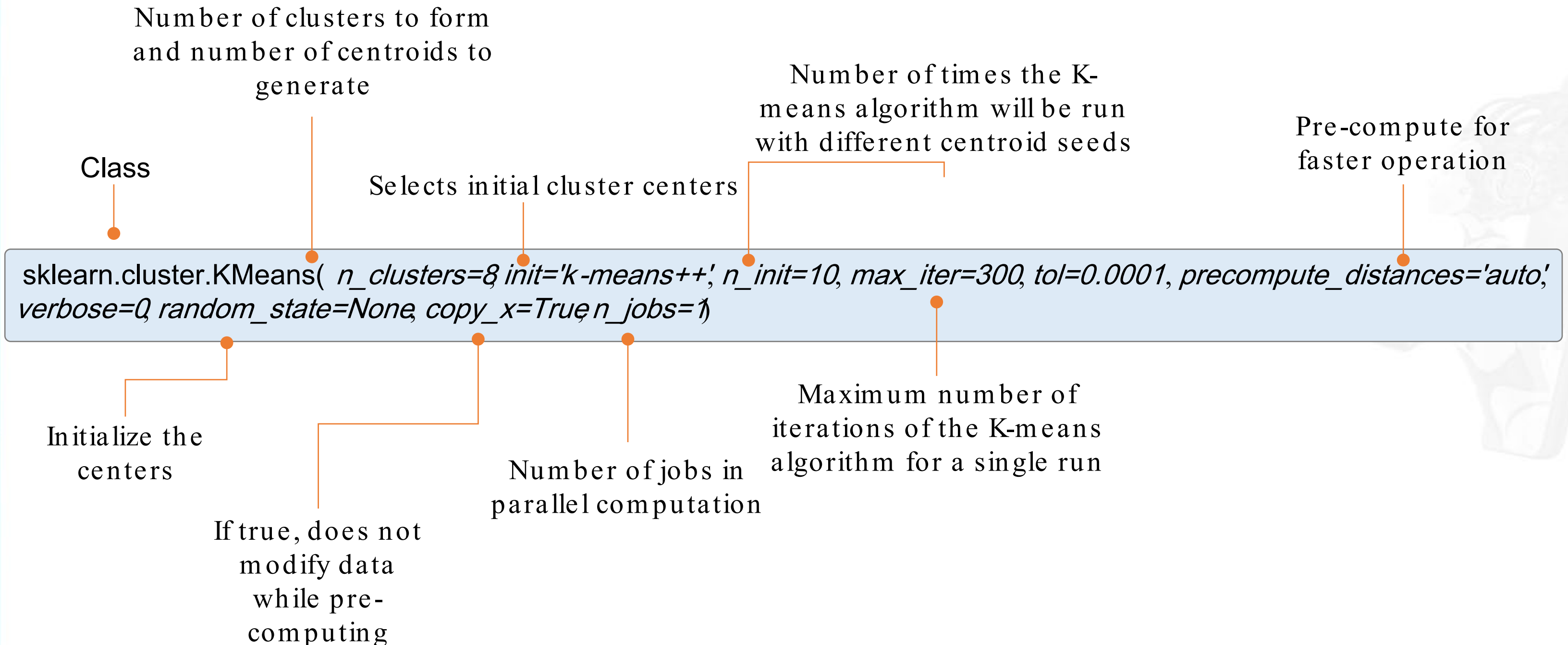
Unsupervised Learning Models: K -Means Clustering

Iterate the process till the model is optimized



Unsupervised Learning Models: K -Means Clustering

Let us see how the k -means algorithm works in Scikit -Learn.



K-Means Clustering



Problem Statement: Demonstrate how to use k -means clustering to classify data points.

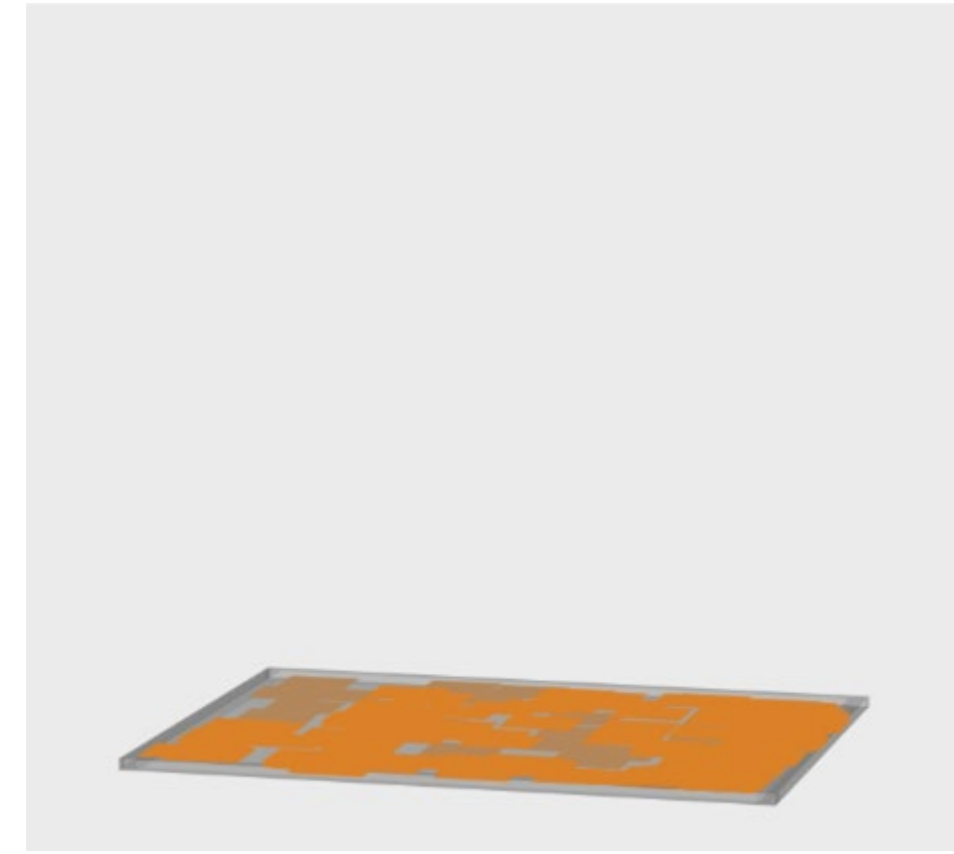
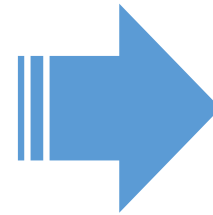
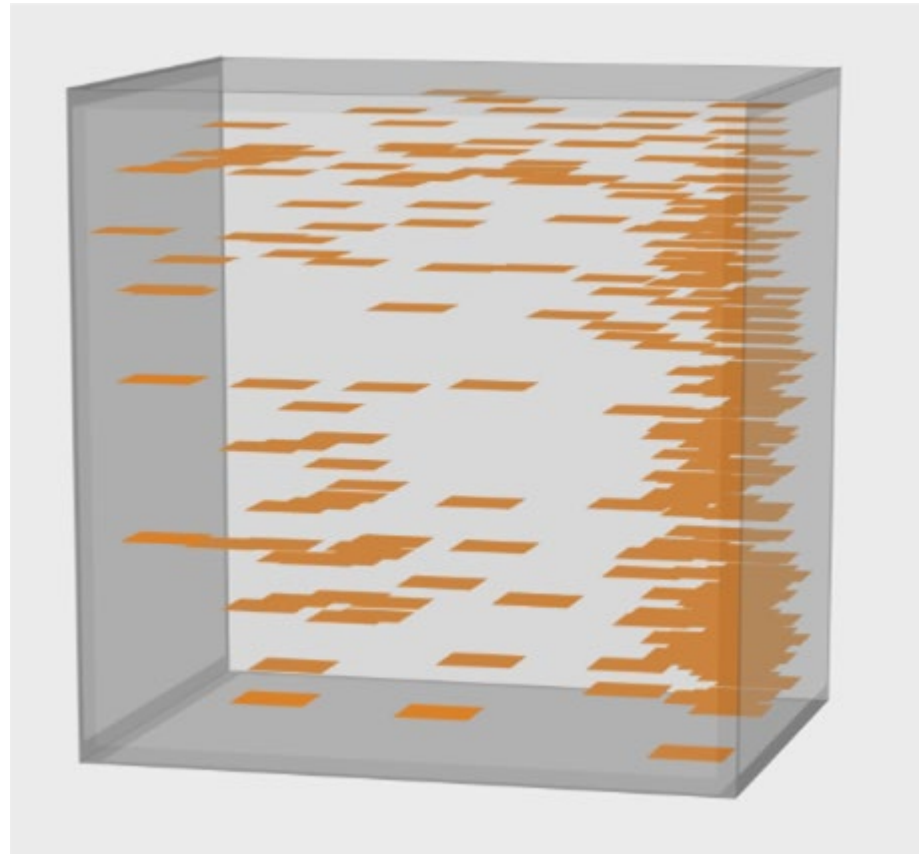
Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login** .

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Unsupervised Learning Models: Dimensionality Reduction

Unsupervised Learning Models: Dimensionality Reduction

It reduces a high -dimensional dataset into a dataset with fewer dimensions. This makes it easier and faster for the algorithm to analyze the data.



Unsupervised Learning Models: Dimensionality Reduction

These are some techniques used for dimensionality reduction:

Drop data columns with missing values

Drop data columns with low variance

Drop data columns with high correlations

Apply statistical functions - PCA

State	Variable	Y1940	Y1941	Y1942	Y1943	Y1944	Y1945	Y1946	Y1947	Y1948	Y1949	Y1950	Y1951	Y1952	Y1953	Y1954	Y1955	Y1956	Y1957	Y1958	Y1959
Alabama	GDP	0.442658	0.315485	0.109414	0.411838	0.87309	0.515454	0.34278	0.719349	0.730522	0.690636	0.721098	0.853162	0.757374	0.18389	0.037726	0.916111	0.889088	0.653694	0.628792	0.291581
Alabama	Unemp	0.652004	0.171815	0.929798	0.878353	0.219275	0.627585	0.126516	0.350107	0.549127	0.101297	0.281344	0.031583	0.478646	0.150707	0.280645	0.639715	0.574256	0.830957	0.244628	0.095407
Alabama	House Price	0.818003	0.2169	0.487288	0.155818	0.048193	0.767306	0.638721	0.760013	0.3129	0.407846	0.33256	0.629126	0.194747	0.672954	0.224241	0.488456	0.829364	0.054417	0.143586	0.917183
Alabama	Prisons	0.656364	0.652592	0.946148	0.352217	0.349515	0.47728	0.372602	0.546885	0.181709	0.424229	0.702031	0.756688	0.969169	0.31284	0.550449	0.411795	0.715701	0.977893	0.057315	0.531458
Alabama	School Budget	0.578769	0.90547	0.277085	0.417637	0.135966	0.990046	0.556206	0.911726	0.020011	0.962333	0.304554	0.690279	0.937539	0.660458	0.648367	0.712692	0.466863	0.452714	0.435571	0.782897
Alabama	Police Budget	0.745779	0.159693	0.150518	0.310762	0.951223	0.583835	0.200034	0.980323	0.574826	0.37626	0.626359	0.818356	0.861317	0.297461	0.435531	0.600432	0.177869	0.576539	0.151442	0.665878
Alaska	GDP	0.511315	0.239402	0.690371	0.902755	0.399771	0.835934	0.120926	0.654525	0.607981	0.118032	0.608898	0.761852	0.159796	0.783235	0.274213	0.842622	0.597706	0.642267	0.819191	0.356878
Alaska	Unemp	0.047899	0.329383	0.039415	0.261495	0.1598	0.518984	0.482901	0.518552	0.606934	0.180587	0.739316	0.124724	0.23812	0.162859	0.318863	0.351793	0.616952	0.018504	0.938324	0.830248
Alaska	House Price	0.430723	0.457658	0.688461	0.359665	0.023767	0.551394	0.35789	0.016201	0.714771	0.116944	0.269702	0.573871	0.383761	0.123655	0.610437	0.63696	0.097339	0.500895	0.664014	0.73167
Alaska	Prisons	0.031685	0.376072	0.836634	0.534889	0.757174	0.848412	0.63369	0.980203	0.0352	0.771676	0.005172	0.311459	0.839973	0.94197	0.898305	0.125163	0.693124	0.79221	0.121708	0.340075
Alaska	School Budget	0.665003	0.59451	0.525446	0.998323	0.015439	0.800823	0.357496	0.706765	0.897678	0.171777	0.417916	0.205629	0.58913	0.553033	0.262769	0.382903	0.732349	0.213502	0.792943	0.777728
Alaska	Police Budget	0.461233	0.227956	0.373834	0.794513	0.411374	0.605271	0.377504	0.775255	0.386393	0.364047	0.858111	0.766152	0.611729	0.901635	0.839296	0.014802	0.774905	0.491703	0.515778	0.866093
Arkansas	GDP	0.751544	0.405693	0.252975	0.160238	0.880534	0.550693	0.950961	0.723004	0.661145	0.503959	0.042729	0.770881	0.735002	0.803669	0.089623	0.336397	0.58918	0.34934	0.61931	0.309744
Arkansas	Unemp	0.592559	0.467335	0.360323	0.411224	0.138199	0.976092	0.113724	0.513229	0.189284	0.962739	0.316513	0.935118	0.414572	0.338363	0.438928	0.415867	0.374687	0.484191	0.338287	0.656182
Arkansas	House Price	0.605316	0.294336	0.824262	0.488503	0.166813	0.336631	0.505568	0.784936	0.771634	0.085395	0.157045	0.25312	0.717742	0.206334	0.519633	0.984801	0.516126	0.125058	0.582149	0.505598
Arkansas	Prisons	0.826649	0.613368	0.100426	0.197751	0.890602	0.804808	0.587972	0.049476	0.586825	0.422374	0.830109	0.458946	0.79477	0.942427	0.503337	0.130984	0.347348	0.249541	0.740134	0.270251
Arkansas	School Budget	0.038493	0.358908	0.020196	0.737979	0.083982	0.270762	0.725234	0.805816	0.552459	0.25004	0.251484	0.840295	0.824462	0.856431	0.397205	0.238459	0.597587	0.84788	0.118996	0.010153
Arkansas	Police Budget	0.362473	0.253577	0.393876	0.126115	0.464451	0.159951	0.887728	0.256411	0.520673	0.263021	0.17562	0.674873	0.826228	0.237534	0.651357	0.604514	0.853966	0.93456	0.251626	0.428394
California	GDP	0.345622	0.033706	0.819983	0.566513	0.011183	0.08491	0.525814	0.396244	0.369756	0.509936	0.464447	0.416412	0.61231	0.409514	0.719096	0.559793	0.26001	0.983729	0.147605	0.359118
California	Unemp	0.744424	0.606742	0.960209	0.025472	0.54712	0.081072	0.368161	0.611313	0.198333	0.80215	0.446168	0.164937	0.225285	0.217991	0.227763	0.386794	0.616075	0.864125	0.949144	0.216368
California	House Price	0.938886	0.231285	0.89848	0.941672	0.383362	0.893858	0.755088	0.43552	0.138745	0.649539	0.087372	0.89598	0.281828	0.512458	0.122694	0.480292	0.640951	0.852	0.930587	0.853051
California	Prisons	0.063822	0.838187	0.94554	0.314737	0.426778	0.894835	0.459681	0.27548	0.988054	0.759395	0.40769	0.154204	0.264051	0.461309	0.588464	0.633898	0.871969	0.808142	0.517105	0.379307
California	School Budget	0.999826	0.884333	0.833853	0.826547	0.598605	0.699173	0.347375	0.498754	0.049039	0.87982	0.129046	0.662399	0.41437	0.470512	0.824108	0.194904	0.944492	0.53874	0.396894	0.639648
California	Police Budget	0.785887	0.272413	0.720612	0.128347	0.314218	0.205528	0.987516	0.013643	0.297284	0.768101	0.729212	0.63841	0.27172	0.409921	0.658117	0.70629	0.104401	0.447338	0.704501	0.325252
Colorado	GDP	0.834676	0.333818	0.602307	0.249385	0.636724	0.473802	0.305362	0.967554	0.83367	0.954399	0.360446	0.748893	0.342564	0.288462	0.653266	0.515591	0.841047	0.389417	0.4379	0.934883
Colorado	Unemp	0.399318	0.516526	0.921863	0.258235	0.047001	0.890816	0.480474	0.75262	0.203198	0.434647	0.521712	0.576462	0.007416	0.979339	0.712837	0.23193	0.129984	0.86499	0.841484	0.828735
Colorado	House Price	0.254889	0.297472	0.99906	0.327953	0.88842	0.219316	0.177885	0.177885	0.193422	0.833027	0.018887	0.152123	0.393048	0.558971	0.088341	0.462896	0.639126	0.635375	0.063241	0.465831
Colorado	Prisons	0.229969	0.788499	0.349294	0.87644	0.230889	0.091053	0.022775	0.960305	0.869051	0.5989	0.700384	0.199959	0.527465	0.788462	0.000822	0.527564	0.979511	0.256275	0.97329	0.183532
Colorado	School Budget	0.391953	0.108194	0.448119	0.512258	0.473698	0.714631	0.967813	0.940385	0.266813	0.202187	0.949271	0.937988	0.599874	0.112504	0.587475	0.065668	0.144408	0.209545	0.244812	
Colorado	Police Budget	0.707277	0.352254	0.38925	0.415637	0.790027	0.278398	0.1078	0.564818	0.174019	0.545961	0.037039	0.279656	0.02462	0.668735	0.672761	0.988568	0.81661	0.801312	0.174024	0.746536
Connecticut	GDP	0.517788	0.309605	0.458049	0.724913	0.159349	0.608148	0.277353	0.245646	0.222746	0.911456	0.93889	0.276782	0.230099	0.214008	0.291571	0.765251	0.049341	0.900581	0.676196	0.817093
Connecticut	Unemp	0.664952	0.955354	0.416259	0.858987	0.49494	0.902229	0.034342	0.001241	0.6417	0.785751	0.561189	0.590625	0.677602	0.056229	0.730139	0.821397	0.799611	0.146413	0.217583	0.442142
Connecticut	House Price	0.482593	0.127942	0.983457	0.398822	0.113141	0.546851	0.112491	0.721477	0.800446	0.479398	0.786782	0.386406	0.250867	0.754481	0.938764	0.635514	0.767172	0.66595	0.431515	0.103553
Connecticut	Prisons	0.114148	0.286225	0.212123	0.346345	0.247112	0.126779	0.394641	0.072685	0.410335	0.878136	0.598487	0.980609	0.817797	0.930455	0.347727	0.012637	0.647032	0.204132	0.072844	0.788484
Connecticut	School Budget	0.789749	0.564194	0.147476	0.433754	0.795206	0.416315	0.364489	0.001990	0.362809	0.115495	0.156211	0.671316	0.725045	0.259556	0.956121	0.780968	0.297336	0.064742	0.183975	0.755835
Connecticut	Police Budget	0.319689	0.093134	0.035506	0.333452	0.590161	0.324209	0.082965	0.812289	0.610024	0.800949	0.20677	0.161142	0.211589	0.282181	0.49841	0.221523	0.139255	0.311718	0.77749	0.884092
Delaware	GDP	0.602571	0.763997	0.023163	0.693981	0.205454	0.946031	0.846575	0.920782	0.178423	0.729485	0.061441	0.69576	0.74773	0.969678	0.553134	0.829911	0.282054	0.460204	0.016191	0.885307
Delaware	Unemp	0.840734	0.121088	0.939174	0.700101	0.129102	0.592598	0.160576	0.994616	0.356692	0.207753	0.967923	0.384214	0.879329	0.106836	0.177633	0.813264	0.281128	0.826093	0.271706	0.662141
Delaware	House Price	0.4554	0.918115	0.335866	0.253401	0.034366	0.604834	0.392335	0.29955	0.873362	0.09431	0.137305	0.590373	0.540046	0.689592	0.662029	0.382379	0.857099	0.758383	0.466922	0.289911
Delaware	Prisons	0.498385	0.206088	0.769942	0.775262	0.449281	0.874381	0.083813	0.456216	0.500752	0.592507	0.895472	0.568386	0.028552	0.730916	0.363413	0.74461	0.342364	0.396626	0.141251	0.506718
Delaware	School Budget	0.906415	0.239993	0.614372	0.740584	0.793066	0.537931	0.933168	0.360957	0.649381	0.265442	0.693759	0.883166	0.099112	0.745121	0.453134	0.622109	0.455377	0.863214	0.020157	0.916889
Delaware	Police Budget	0.065336	0.938448	0.597192	0.275946	0.320259	0.75084	0.032236	0.199282	0.504855	0.666764	0.613526	0.840203	0.215207	0.104328	0.520208	0.811331	0.573414	0.353654	0.218152	0.5218
Florida	GDP	0.659796	0.567298	0.102078	0.970927	0.473231	0.835698	0.003362	0.148081	0.881319	0.947536	0.519067	0.80723	0.542673	0.969883	0.136805	0.121218	0.171623	0.458686	0.384353	0.670016

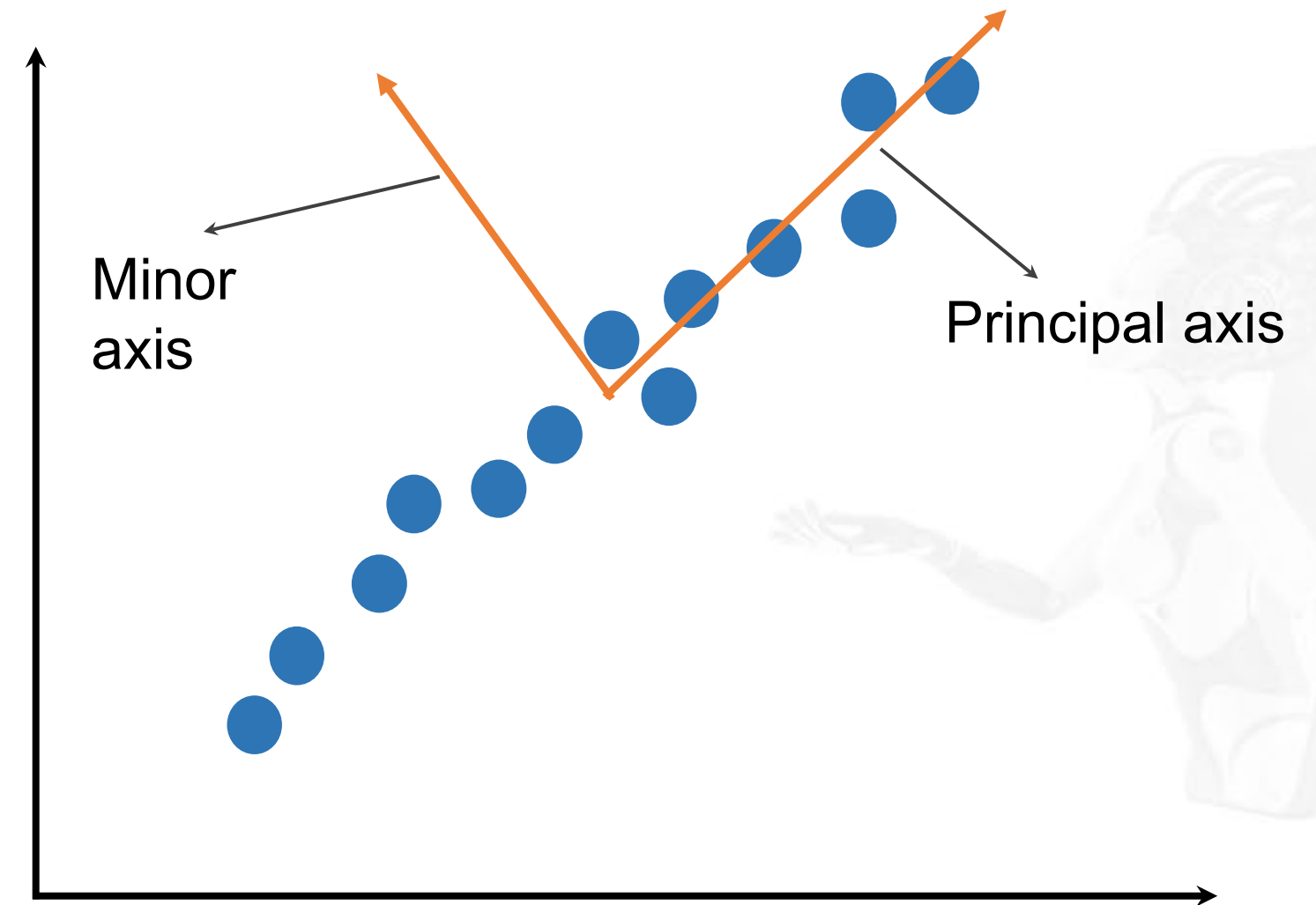
Large dataset
(a few thousand columns and rows)

Unsupervised Learning Models: Principal Component Analysis

Unsupervised Learning Models: Principal Component Analysis (PCA)

It is a linear dimensionality reduction method which uses singular value decomposition of the data and keeps only the most significant singular vectors to project the data to a lower dimensional space.

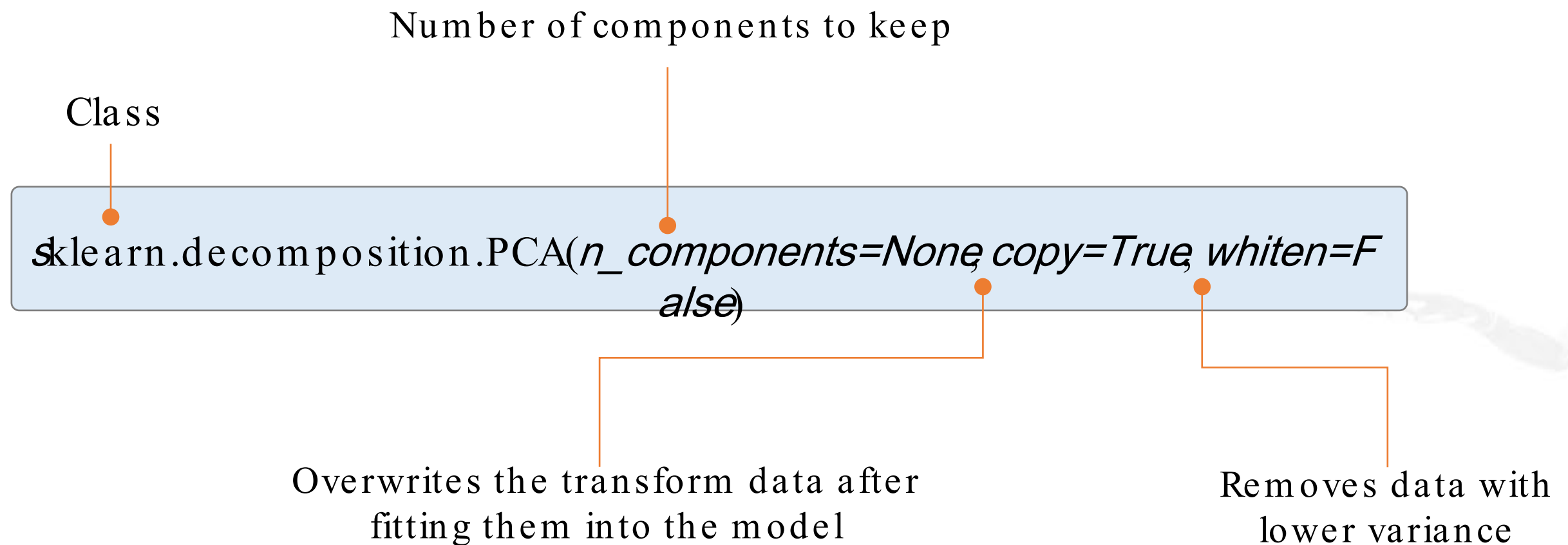
- It is primarily used to compress or reduce the data.
- PCA tries to capture the variance, which helps it pick up interesting features.
- PCA is used to reduce dimensionality in the dataset and to build our feature vector.
- Here, the principal axes in the feature space represents the direction of maximum variance in the data.



This method is used to capture variance.

Unsupervised Learning Models: Principal Component Analysis

Let us look at how the PCA algorithm works in Scikit -Learn.



Principal Component Analysis (PCA)

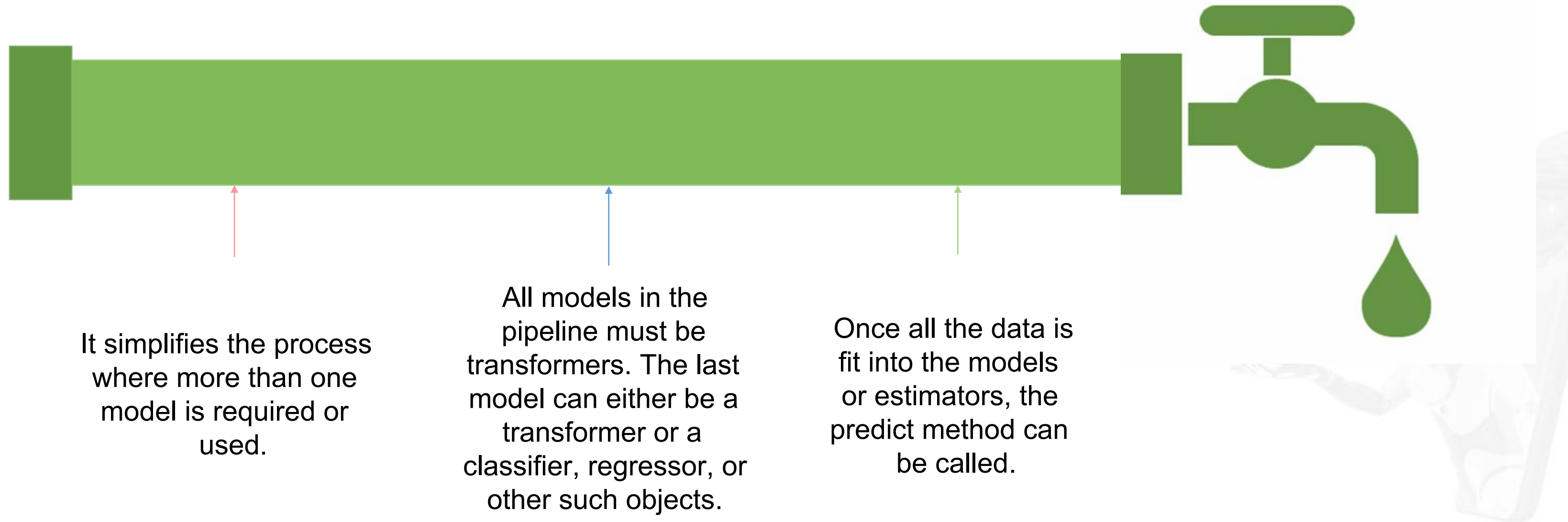


Problem Statement: Demonstrate how to use the PCA model to reduce the dimensions of a dataset.

Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login**.

ASSISTED PRACTICE

Pipeline



Estimators are known as 'model instance'.

Pipelines



Problem Statement: Demonstrate how to build a pipeline.

Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login**.

ASSISTED PRACTICE

Model Persistence

You can save your model for future use. This avoids the need to retrain the model.

- This can be saved using the Pickle method.
- It can also be replaced with the joblib of Sci-kit team.
- Both `joblib.dump` and `joblib.load` can be used.
- These would be efficient for Big Data.



Model Persistence



Problem Statement: Demonstrate how to persist a model for future use.

Access: Click on the **Practice Labs** tab on the left side panel of the LMS. Copy or note the username and password that is generated. Click on the **Launch Lab** button. On the page that appears, enter the username and password in the respective fields, and click **Login**.

ASSISTED PRACTICE

Model Evaluation: Metric Functions

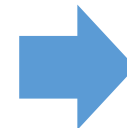
You can use the “Metrics” function to evaluate the accuracy of your model’s predictions.

Classification



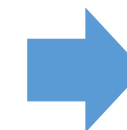
```
metrics.accuracy_score  
metrics.average_precision_score
```

Clustering



```
metrics.adjusted_rand_score
```

Regression



```
metrics.mean_absolute_error  
metrics.mean_squared_error  
metrics.median_absolute_error
```



Evaluation of Dataset



Problem Statement:

The given dataset contains ad budgets for different media channels and the corresponding ad sales of XYZ firm. Evaluate the dataset to:

- Find the features or media channels used by the firm
- Find the sales figures for each channel
- Create a model to predict the sales outcome
- Split as training and testing datasets for the model
- Calculate the Mean Square Error (MSE)

UNASSISTED PRACTICE

Evaluation of Dataset



Instructions to perform the assignment:

- Download the FAA dataset from the “Resource” tab. Upload the dataset to your Jupyter notebook to view and evaluate it.

Common instructions:

- If you are new to Python, download the “Anaconda Installation Instructions” document from the “Resources” tab to view the steps for installing Anaconda and the Jupyter notebook.
- Download the “Assignment 01” notebook and upload it on the Jupyter notebook to access it.
- Follow the cues provided to complete the assignment.

UNASSISTED PRACTICE

Listing the Glucose Level Readings



Problem Statement:

The given dataset lists the glucose level readings of several pregnant women taken either during a survey examination or routine medical care. It specifies if the two hours post load plasma glucose was at least 200 mg/dl. Analyze the dataset to:

1. Find the features of the dataset
2. Find the response label of the dataset
3. Create a model to predict the diabetes outcome
4. Use training and testing datasets to train the model
5. Check the accuracy of the model

UNASSISTED PRACTICE

Listing the Glucose Level Readings



Instructions on performing the assignment:

- Download the “pima -Indians -diabetes.DATA” and “pima -Indians -diabetes.NAMES” files from the “Resources” tab. Load the .DATA file to the Jupyter notebook to work on it.
- Open the .NAMES file with a notepad application to view its text. Use this file to view the features of the dataset and add them manually in your code.

Common instructions:

- If you are new to Python, download the “Anaconda Installation Instructions” document from the “Resources” tab to view the steps for installing Anaconda and the Jupyter notebook.
- Download the “Assignment 02” notebook and upload it on the Jupyter notebook to access it.
- Follow the provided cues to complete the assignment.

Key Takeaways

You are now able to:

- 🕒 Define machine learning
- 🕒 Explain the machine learning approach
- 🕒 List relevant terminologies that help you understand a dataset
- 🕒 Discuss features of supervised and unsupervised learning models
- 🕒 Explain algorithms, such as regression, classification, clustering, and dimensionality reduction



DATA AND ARTIFICIAL INTELLIGENCE



Knowledge Check

Knowledge Check

1

In machine learning, which one of the following is an observation?

- a. Features
- b. Attributes
- c. Records
- d. Labels

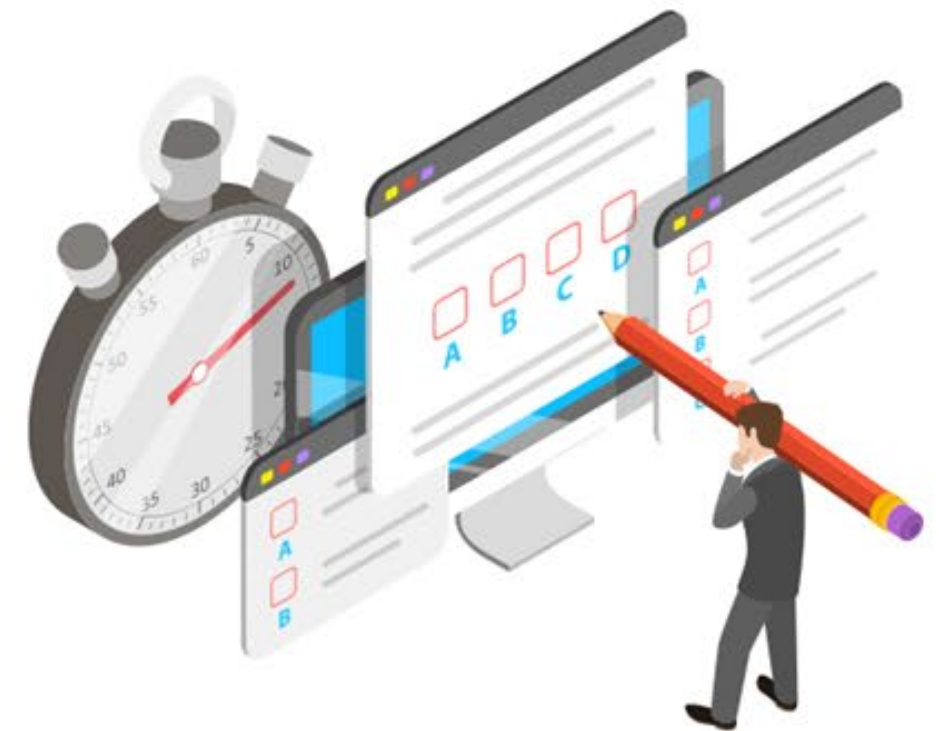


Knowledge Check

1

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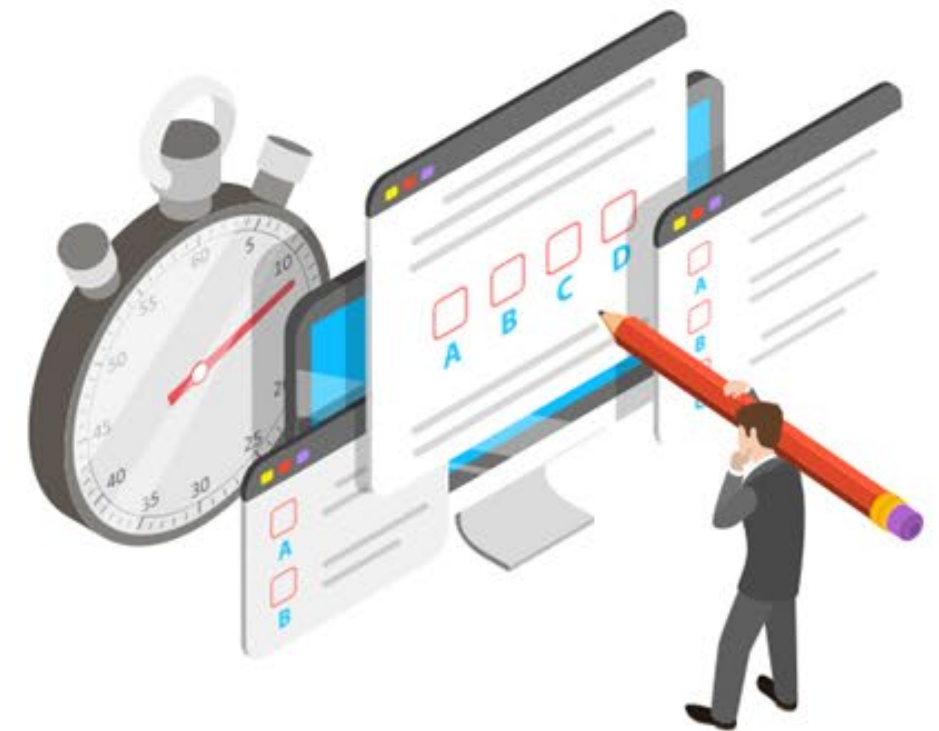


The correct answer is **C**

An observation is a set of examples, records, or samples.

If data is continuous and has labels (response), then it fits which of the following problem types?

- a. Supervised learning: Classification
- b. Unsupervised learning: Clustering
- c. Unsupervised learning: Dimensionality reduction
- d. Supervised learning: Regression



Knowledge
Check

2

If data is continuous and has labels (response), then it fits which of the following problem types?

- a. Supervised learning: Classification
- b. Unsupervised learning: Clustering
- c. Unsupervised learning: Dimensionality reduction
- d. Supervised learning: Regression



The correct answer is **d**

The regression algorithm belonging to the supervised learning model is best suited to analyze continuous data.

Knowledge Check

3

Identify the goal of unsupervised learning. Select all that apply.

- a. To predict the outcome
- b. To understand the structure of the data
- c. To generalize the dataset
- d. To represent the data

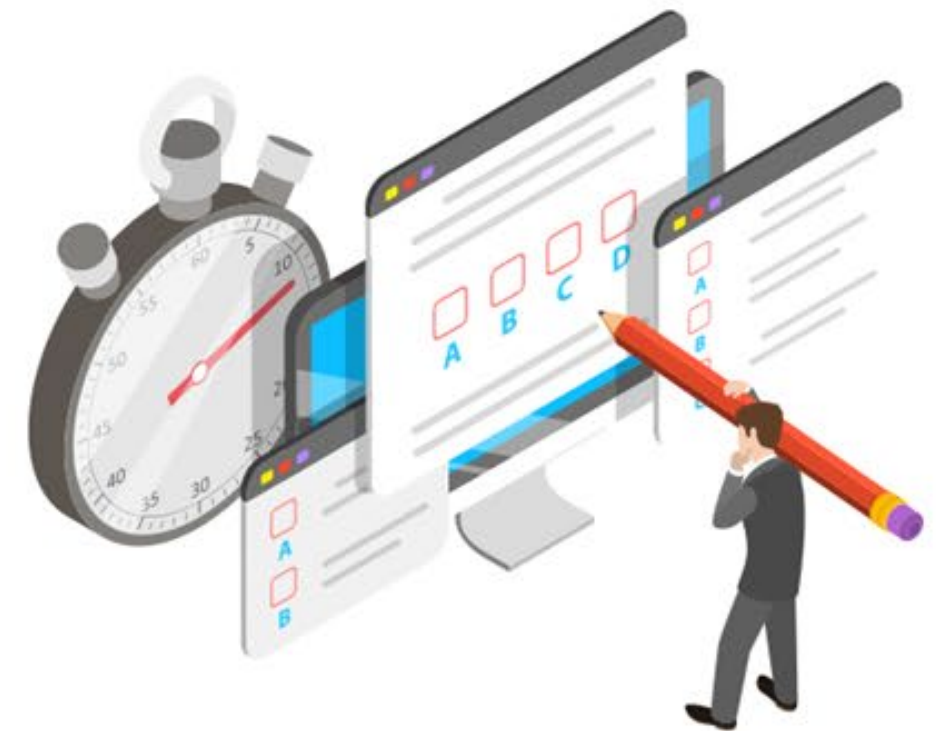


Knowledge Check

3

Identify the goal of unsupervised learning. Select all that apply.

- a. To predict the outcome
- b. To understand the structure of the data
- c. To generalize the dataset
- d. To represent the data



The correct answer is **b, d**

The goal of unsupervised learning is to understand the structure of the data and represent it. There is no right or certain answer in unsupervised learning.

The estimator instance in Scikit -learn is a _____.

- a. To predict the outcome
- b. To understand the structure of the data
- c. To generalize the dataset
- d. To represent the data

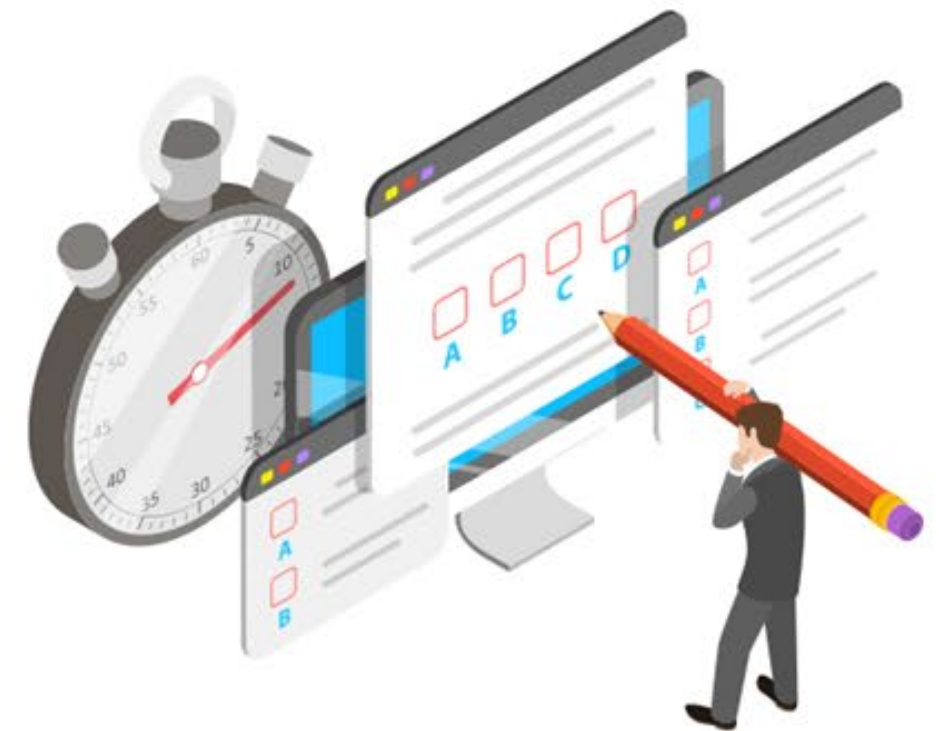


Knowledge
Check

4

The estimator instance in Scikit -learn is a _____.

- a. Model
- b. Feature
- c. Dataset
- d. Response

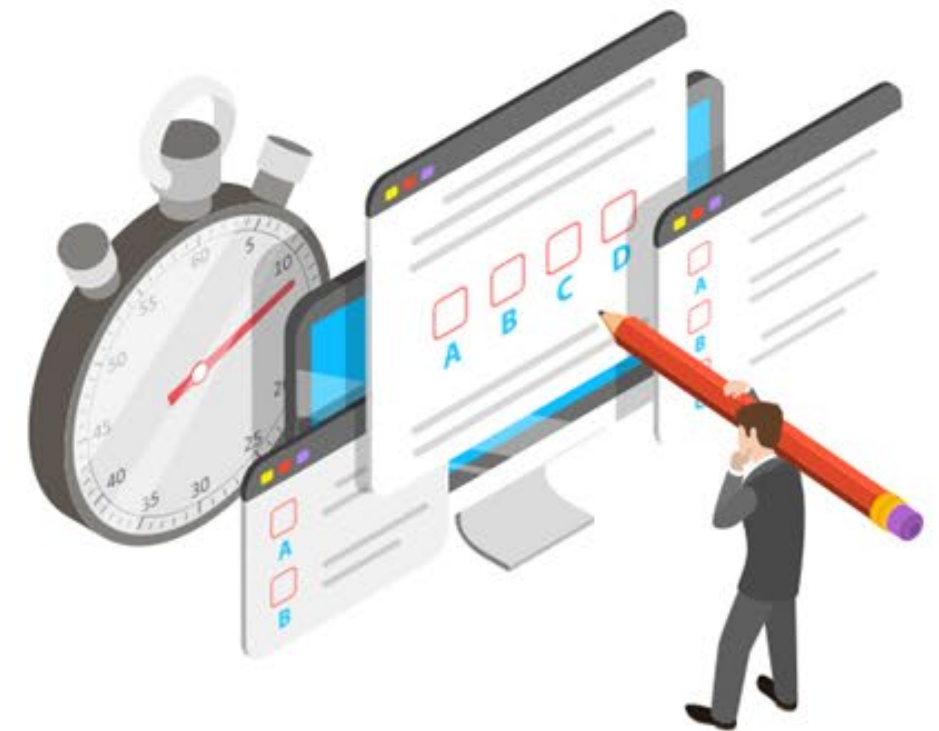


The correct answer is **a**

The estimator instance or object is a model.

What is the best way to train a model?

- a. Use the entire dataset as a training and testing set both
- b. Split the known dataset into separate training and testing sets
- c. Ask the source to provide continuous data
- d. Ask the source to provide categorical data



What is the best way to train a model?

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The correct answer is **b**

The best way to train a model is to split the known dataset into training and testing sets. The testing set varies from 20% to 40%.

Which of the following is true with a greater value of SSR or SSE? Select all that apply.

- a. The prediction will be more accurate, making it the best fit model.
- b. The prediction will start becoming less accurate.
- c. The outcome remains unaffected.
- d. The model will not be the best fit for the attributes.

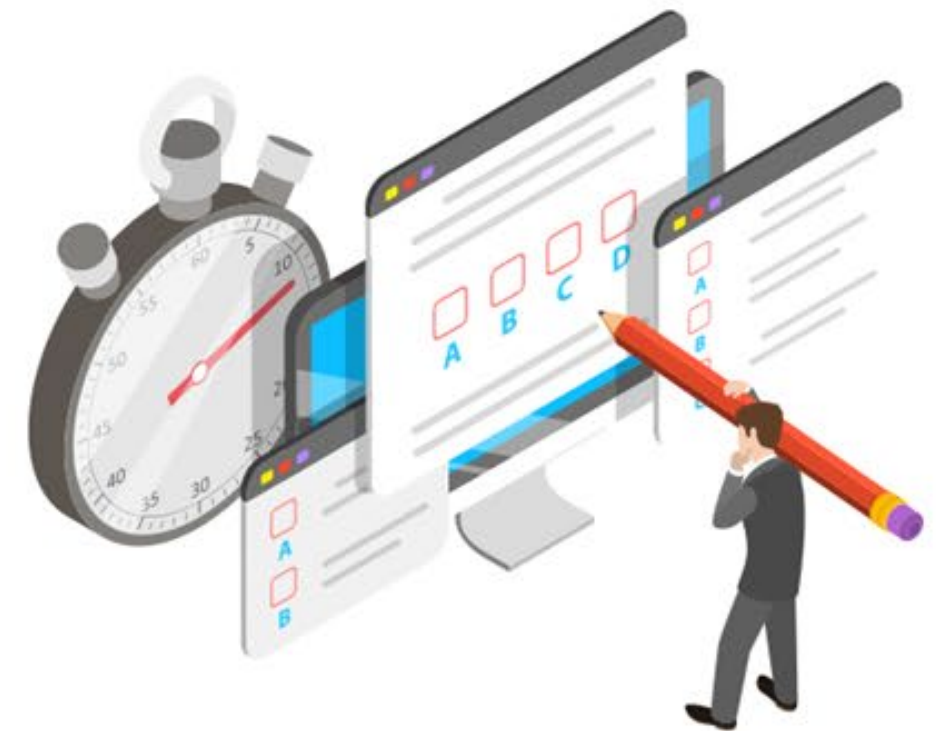


Knowledge
Check

6

Which of the following is true with a greater value of SSR or SSE? Select all that apply.

- a. The prediction will be more accurate, making it the best fit model.
- b. The prediction will start becoming less accurate.
- c. The outcome remains unaffected.
- d. The model will not be the best fit for the attributes.



The correct answer is **b, d**

With higher SSR or SSE, the prediction will be less accurate and the model will not be the best fit for the attributes.

Class `sklearn.linear_model.LogisticRegression`, `random_state` ____.

- a. Indicates the seed of the pseudo random number generator used to shuffle data
- b. Defines the features state
- c. Represents the number of random iterations
- d. Specifies a random constant to be added to the decision function



Class `sklearn.linear_model.LogisticRegression`, `random_state` _____.

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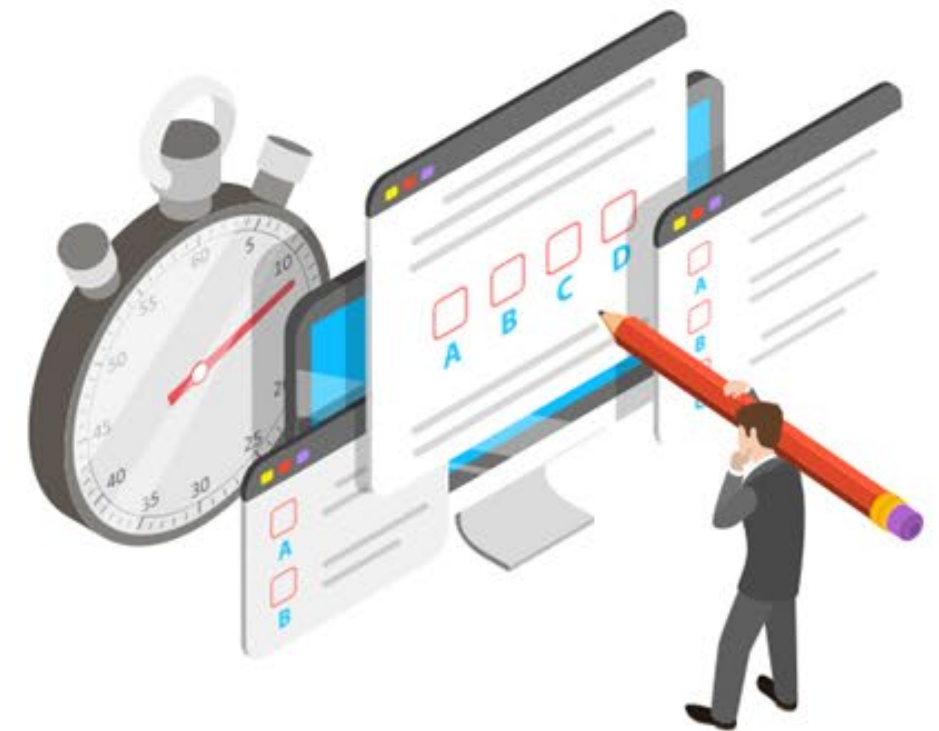


The correct answer is **a**

The class “`sklearn.linear_model.LogisticRegression`, `random_state`” indicates the seed of the pseudo random number generator used to shuffle data.

What are the requirements of the K -means algorithm? Select all that apply.

- a. Number of clusters should be specified
- b. More than one iteration should meet requisite criteria
- c. Centroids should minimize inertia
- d. Features should be labeled



What are the requirements of the K -means algorithm? Select all that apply.

- a. Number of clusters should be specified
- b. More than one iteration should meet requisite criteria
- c. Centroids should minimize inertia
- d. Features should be labeled



The correct answer is **a, b, c**

The K-means algorithm requires the number of clusters to be specified and the centroids to minimize inertia. It requires several iterations to fine tune itself and meet the required criteria to become the best fit model.

What are the requirements of the K -means algorithm? Select all that apply.

- a. Fits the model with X and applies the dimensionality reduction on X
- b. Transforms the data back to its original space
- c. Applies the dimensionality reduction on X
- d. Computes data co-variance with the generative model



In Class `sklearn.decomposition.PCA`, the `transform(X)` method , where X is multi dimensional _____. -

- a. Fits the model with X and applies the dimensionality reduction on X
- b. Transforms the data back to its original space
- c. Applies the dimensionality reduction on X
- d. Computes data co-variance with the generative model



The correct answer is **C**

In Class “`sklearn.decomposition.PCA`,” the `transform(X)` method applies the dimensionality reduction on X.

Thank You