



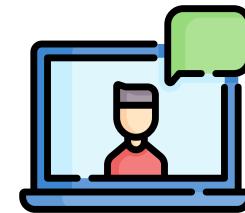
Data Warehouses

Definição da tecnologia



- Armazém de dados
- Coleção de dados orientada a assunto
- Características:
 - Base SQL
 - Pré-processamento
 - Visualização
 - ML/AI
- Otimizada
- Diversas tecnologias
- Data Mart
- Data Lake

Introdução

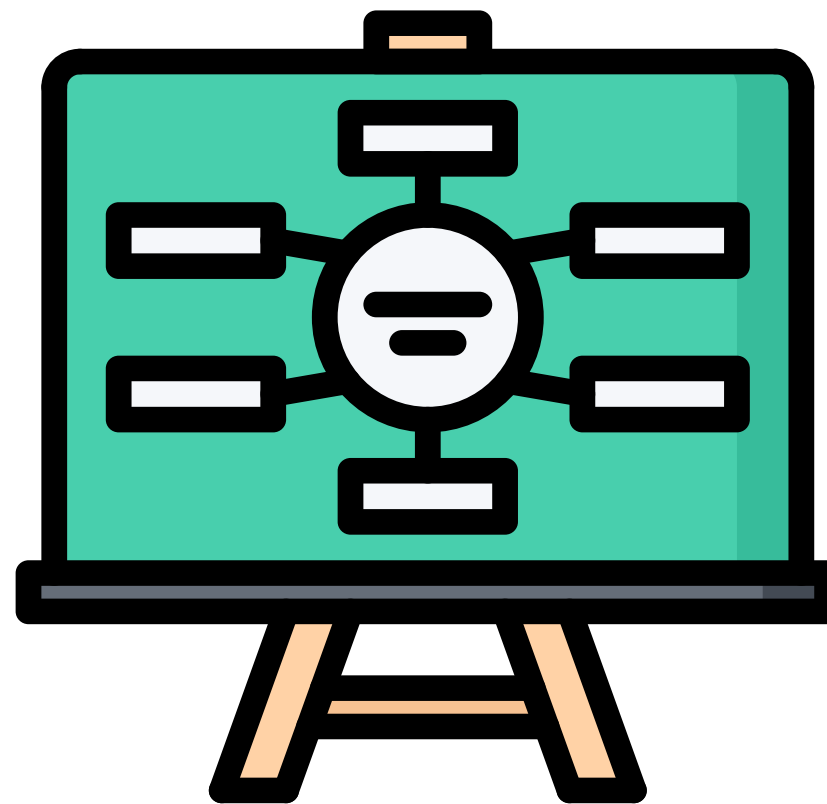


- Base de dados foi encontrada no Kaggle
- Base de dados sobre tweets da eleição dos EUA.
- Base de Dados no formato de arquivo csv
- O porquê essa base de dados foi escolhida

Schema de um data warehouse

Conceitos

- Fact tables que são tabela principal que reúne os atributos das tabelas secundárias mais os seus próprios atributos principais (exemplo: valor total da venda);
- Dimension tables, tabela secundária de características sobre a tabela principal (exemplo: horário que a venda foi realizada).



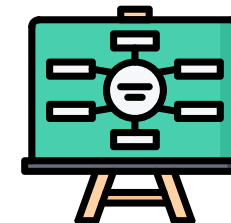
Modelando o Banco



- Análise do CSV
- Modelagem SQL
- Script Python para tratamento dos dados
 - ETL(Extract Transform Load)
 - Pandas
 - SQLAlchemy

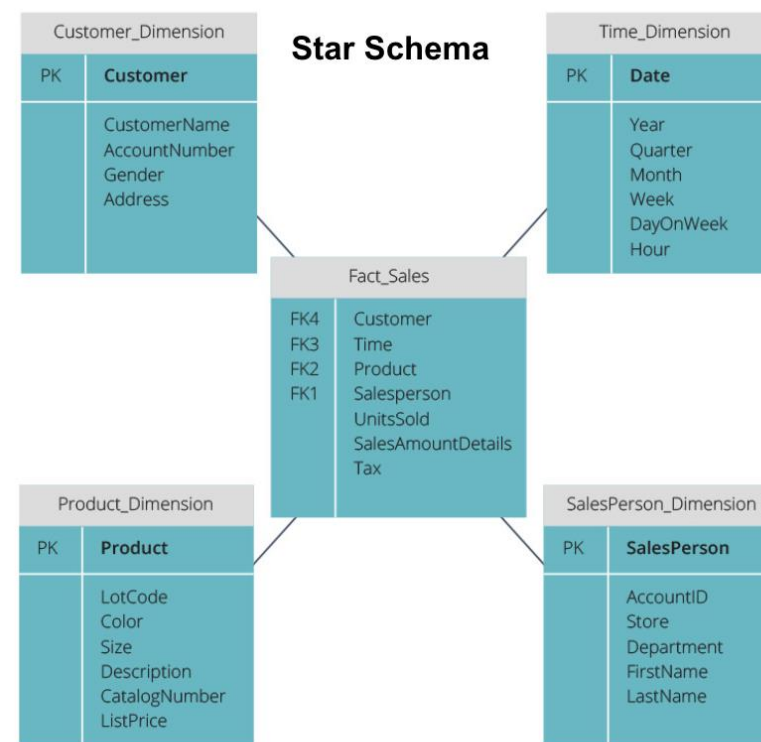


Escolha de um Schema



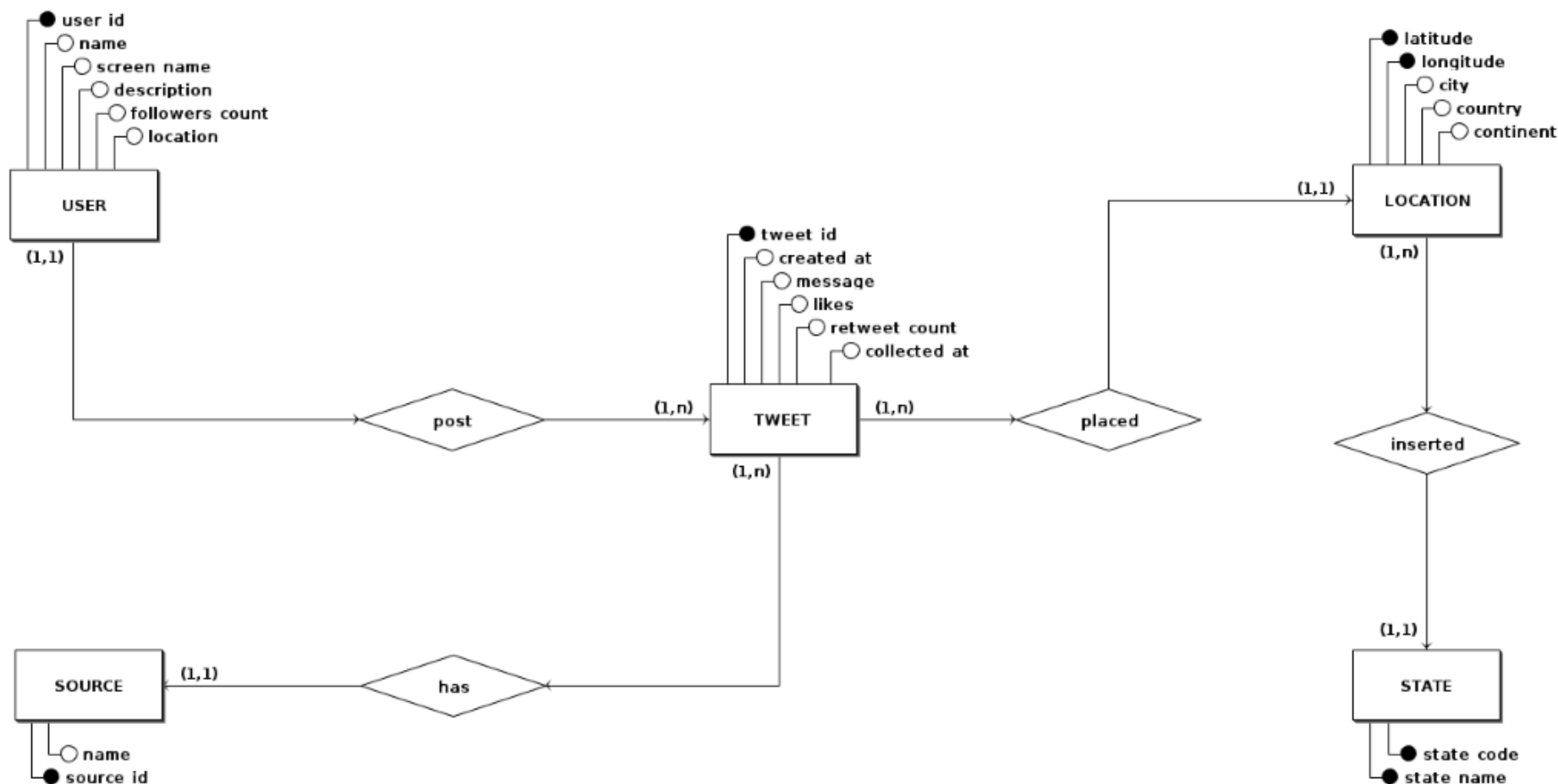
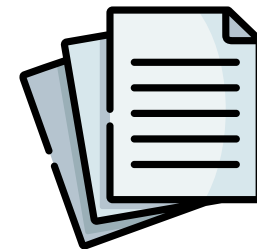
Star Schema

- Dividir fact table em dimension tables desnormalizadas.
- Maior velocidade de consultas por realizar apenas uma junção das dimension tables com a fact table.
- Escolha mais adequada para o contexto



Legend: PK = primary key | FK = foreign key

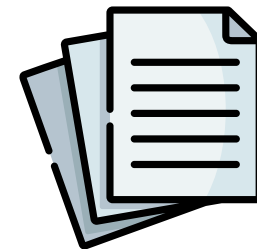
Documentação da Base de Dados



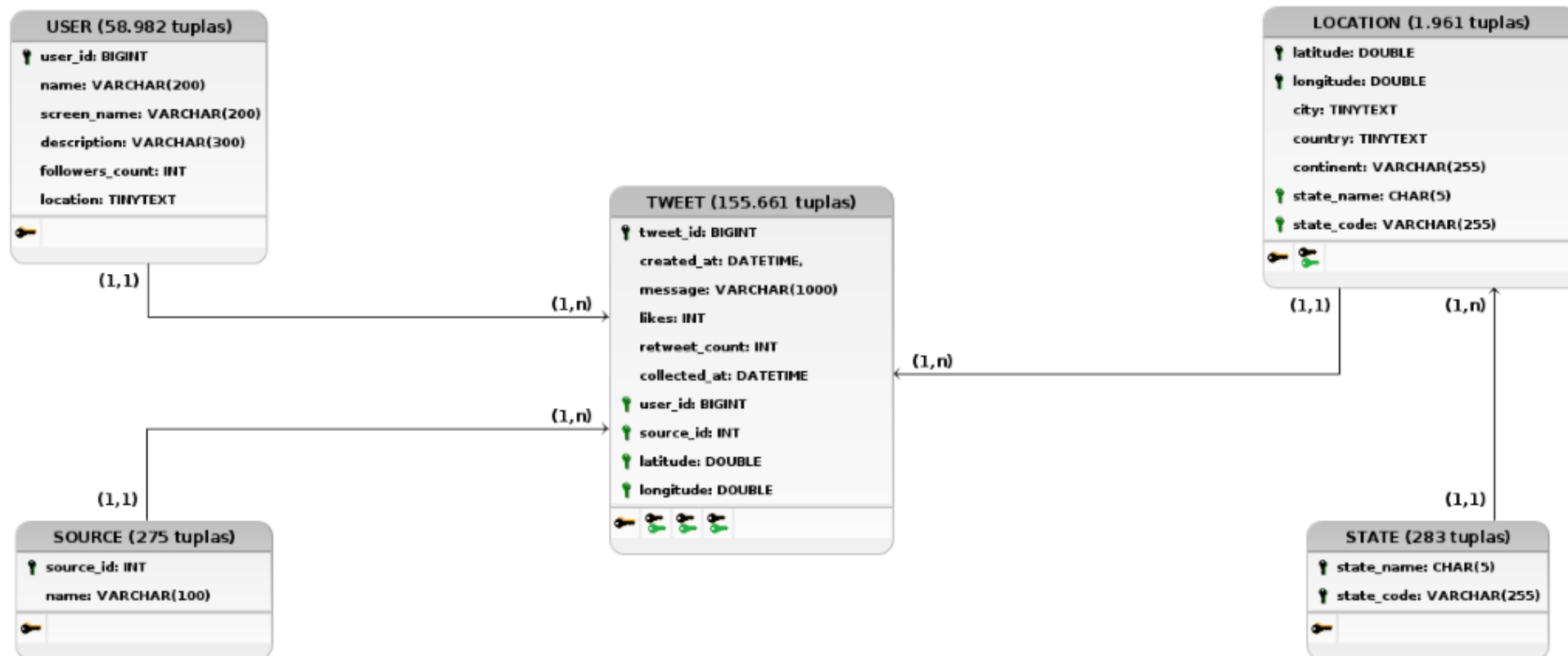
[Kaggle](#)

[Dump](#)

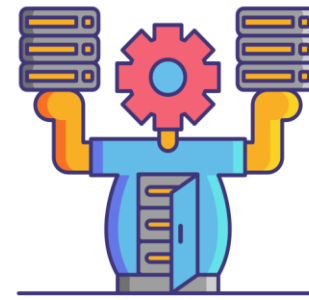
Documentação da Base de Dados



[Kaggle](#)
[Dump](#)



Trabalhando com Big Data



- Dump gerado
- Problemas com exportação de dados (Encoding)
- Inserts do Dump
- Formatação via Regex

Consultas



us_election_2020.FACT_TWEET	
123 tweet_id	bigint NOT NULL
🕒 created_at	datetime
ABC message	varchar(1000) NOT NULL
123 likes	int NOT NULL
123 retweet_count	int NOT NULL
123 source_id	int NOT NULL
123 user_id	bigint NOT NULL
123 latitude	double NOT NULL
123 longitude	double NOT NULL
🕒 collected_at	datetime

us_election_2020.LOCATION_DIMENSION	
123 latitude	double NOT NULL
123 longitude	double NOT NULL
ABC city	tinytext NOT NULL
ABC country	tinytext NOT NULL
ABC continent	varchar(200) NOT NULL
ABC state_code	char(5) NOT NULL
ABC state_name	varchar(255) NOT NULL

us_election_2020.STATE_DIMENSION	
ABC state_code	char(5) NOT NULL
ABC state_name	varchar(255) NOT NULL

us_election_2020.TWEETS_BY_STATE	
ABC state	char(5) NOT NULL
ABC name	varchar(255) NOT NULL
123 quantity	bigint NOT NULL

Consultas



```
SELECT * FROM TWEETS_BY_STATE LIMIT 10
```

	state	name	quantity
1	NY	New york	18,112
2	CA	California	15,935
3	ENG	England	11,786
4	DC	District of columbia	6,703
5	TX	Texas	6,691
6	IDF	Ile-de-france	6,669
7	FL	Florida	5,405
8	IL	Illinois	3,478
9	DL	Delhi	3,477
10	ON	Ontario	3,280

Consultas



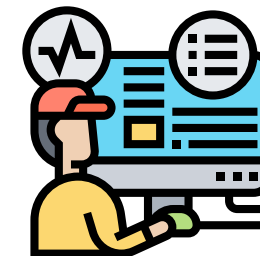
us_election_2020.FACT_TWEET	
123 tweet_id	bigint NOT NULL
🕒 created_at	datetime
ABC message	varchar(1000) NOT NULL
123 likes	int NOT NULL
123 retweet_count	int NOT NULL
123 source_id	int NOT NULL
123 user_id	bigint NOT NULL
123 latitude	double NOT NULL
123 longitude	double NOT NULL
🕒 collected_at	datetime

us_election_2020.USER_DIMENSION	
123 user_id	bigint NOT NULL
ABC name	varchar(200) NOT NULL
ABC screen_name	varchar(200) NOT NULL
ABC description	varchar(300) NOT NULL
123 followers_count	int NOT NULL
ABC location	tinytext NOT NULL



us_election_2020.USERS_ANALYTICS	
123 id	bigint NOT NULL
ABC username	varchar(200) NOT NULL
123 followers	int NOT NULL
123 qtd_tweets	bigint NOT NULL
123 likes_avg	decimal(14,4)
123 retweet_avg	decimal(14,4)

Consultas



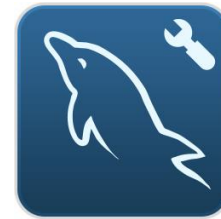
SELECT * FROM USERS_ANALYTICS WHERE likes_avg > 1000000 Enter a SQL expression to filter results (use Ctrl+Space)

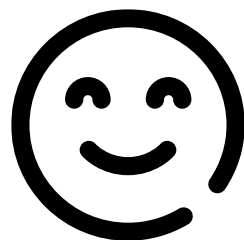
	123 id	abc username	123 followers	123 qtd_tweets	123 likes_avg	123 retweet_avg
1	105,165,371	andrebercoff	88,139	5	2,243.8	797.4
2	550,377,428	AshokShrivasta6	101,742	3	2,407.6667	373.3333
3	19,409,508	bobbyberk	444,512	2	2,854.5	102
4	2,329,492,448	OliviaTroye	118,266	2	2,813	608.5
5	18,208,368	kurtbardella	68,810	2	2,757.5	847
6	509,981,951	PorcherThomas	96,998	2	2,608	595
7	196,401,315	johndelancie	71,438	2	2,127	407
8	394,250,799	Br4mm3n	335,858	2	2,063	14.5
9	1,591,400,760	TaraSetmayer	105,668	2	2,011	178
10	56,039,856	abhijitmajumder	547,842	1	2,797	956
11	87,643,561	Rajput_Ramesh	91,070	1	2,749	231
12	1,066,505,848,791,474,176	FamilyKabs	38,636	1	2,605	352
13	14,085,099	KDKA	193,015	1	2,525	683
14	1,664,075,966	HeatherTDay	38,102	1	2,348	92
15	412,591,867	masonsixtencox	39,062	1	2,299	99
16	393,421,515	BernardineEvari	52,353	1	2,268	99
17	74,169,715	MzKatieCassidy	671,142	1	2,248	139
18	122,104,353	c_lindner	414,994	1	2,199	115
19	14,957,147	TheYoungTurks	429,163	1	2,123	469
20	1,249,274,011,109,949,440	TorreyHarris901	1,508	1	2,035	204

Ferramentas



- DBeaver
- MySQL Workbench
- Vscode
- BrModelo





Obrigado!

