

DATA SCIENTIST WAS WORK FLOW

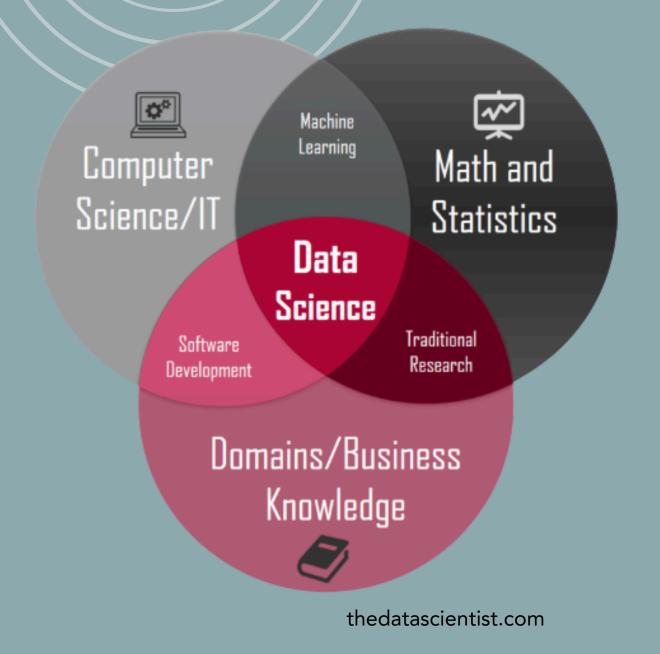
Harish Muhammad

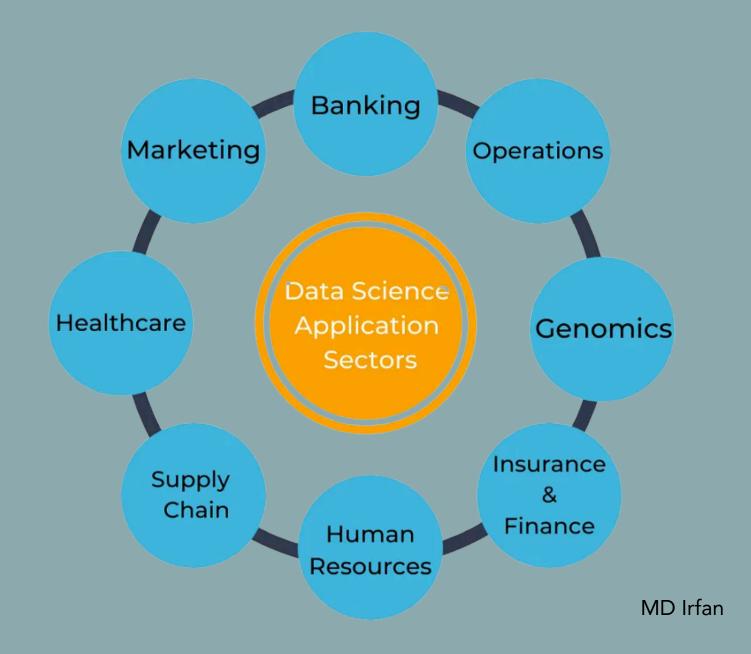
- 1. APA ITU DATA SCIENCE?
- O2 ALUR KERJA DATA SAINS
- O3 PEMAHAMAN BISNIS
- 04 DATA UNDERSTANDING
- 05. DATA PREPARATION
- O6. PEMODELAN MACHINE LEARNING
- 07. EVALUASI
- 08 MODEL DEPLYOMENT
- 09. KESIMPULAN



OUTLINE PRESENTASI

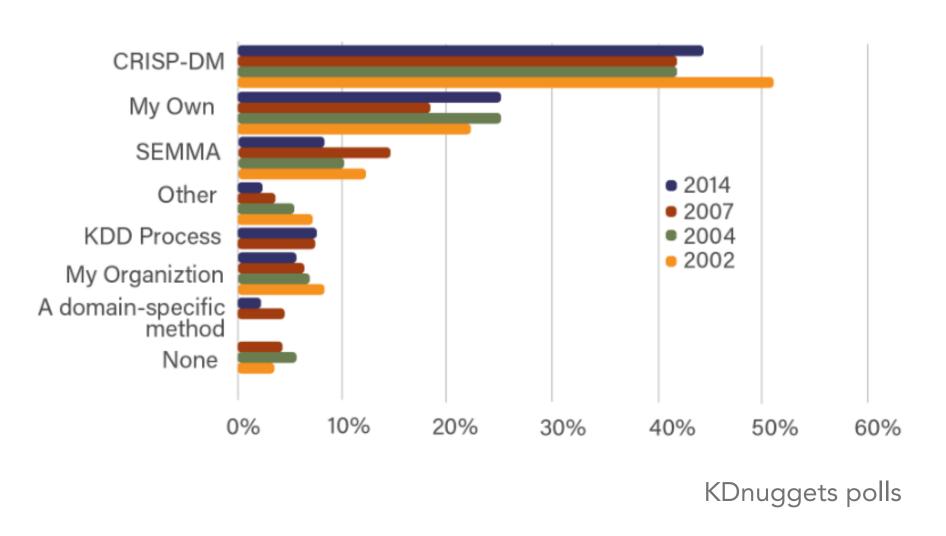
APA ITU DATA SAINS?



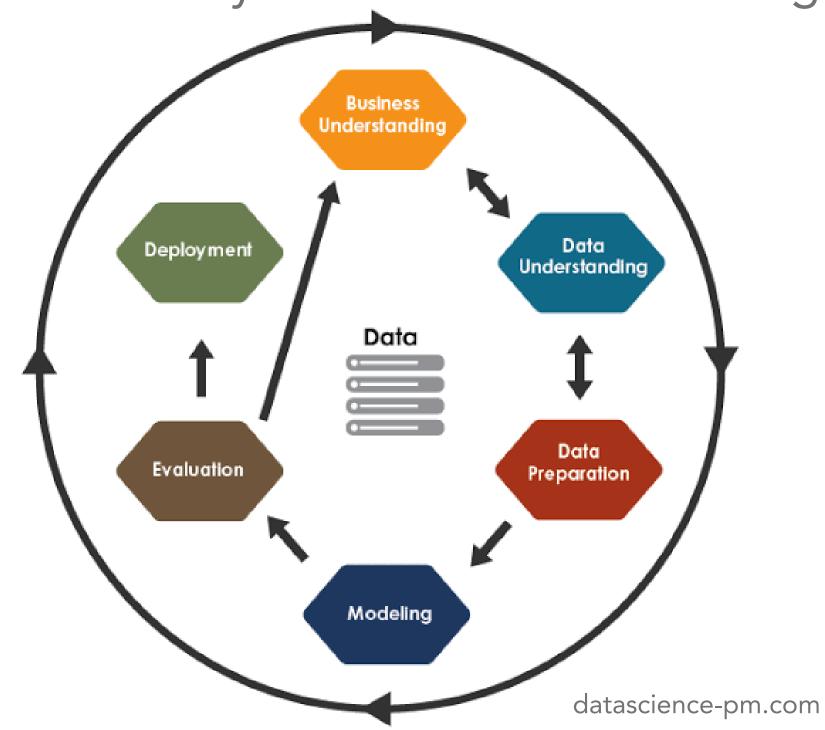


ALUR KERJA BERDASARKAN CRISP-DM

Popularitas Penggunaan CRISP-DM



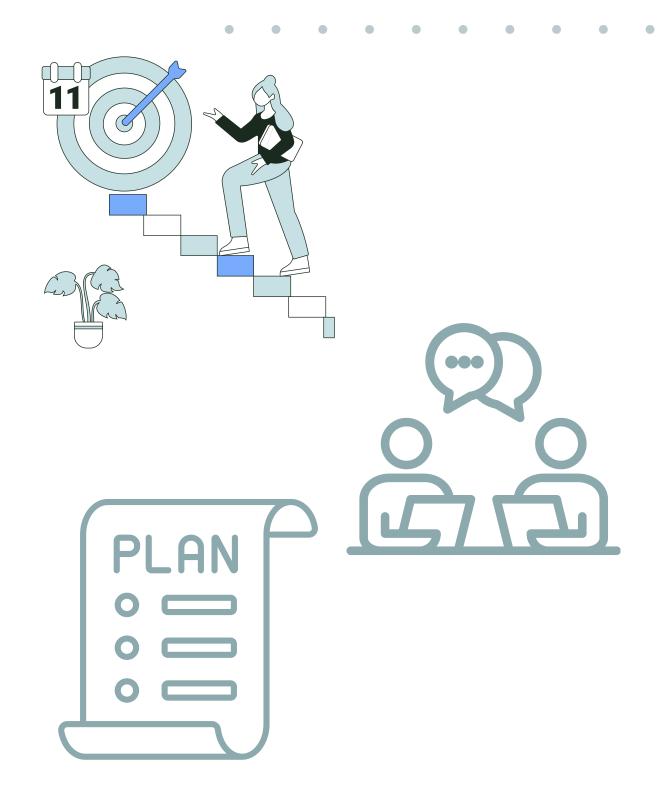
Cross Industry standard for data mining





PEMAHAMAN BISNIS (BUSINESS UNDERSTANDING)

- Memahami orientasi & kebutuhan bisnis
- Berdiskusi dengan stakeholders
- Melihat situasi & ketersediaan sumber daya
- Menentukan tujuan proyek DS
- Perencanaan proyek DS

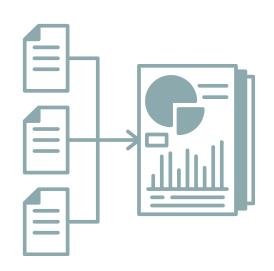


PEMAHAMAN DATA (DATA UNDERSTANDING)



Pengumpulan data

- Sumber data
- Lokasi
- Metode
- Unsur privasi



Data Description

- Tipe data
- Volume data



Explorasi Data

- Visualisasi data
- Statistik deskriptif
- Trend data
- Korelasi



Pengumpulan data, deskripsi dan explorasi data

```
import pandas as pd
file_name = 'diabetes_data.csv'
df = pd.read_csv(file_name)
display(df.info(), df.describe(include='0').T)
```

Visualisasi data

```
import matplotlib.pyplot as plt
df.hist(bins=10, figsize=(10,5))
plt.show()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 520 entries, 0 to 519
Data columns (total 17 columns):
                        Non-Null Count Dtype
    Column
                        _____
                        520 non-null
                                       int64
    Genden
                        520 non-null
                                       object
    Polyuria
                        520 non-null
                                        object
    Polydipsia
                        520 non-null
                                        object
    sudden weight loss 520 non-null
                                        object
    weakness
                        520 non-null
                                        object
    Polyphagia
                        520 non-null
                                       object
    Genital thrush
                        520 non-null
                                        object
    visual blurring
                        520 non-null
                                       object
    Itching
                        520 non-null
                                        object
    Irritability
                        520 non-null
                                        object
    delayed healing
                        520 non-null
                                        object
    partial paresis
                        520 non-null
                                        object
13 muscle stiffness
                        520 non-null
                                        object
14 Alopecia
                        520 non-null
                                        object
15 Obesity
                        520 non-null
                                        object
16 class
                        520 non-null
                                        object
```

dtypes: int64(1), object(16) memory usage: 69.2+ KB None

Age

120

100

80

40

20

30

40

50

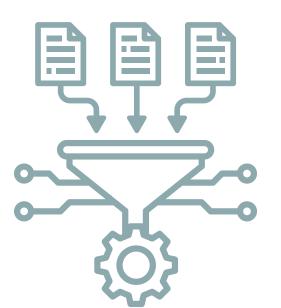
60

70

80

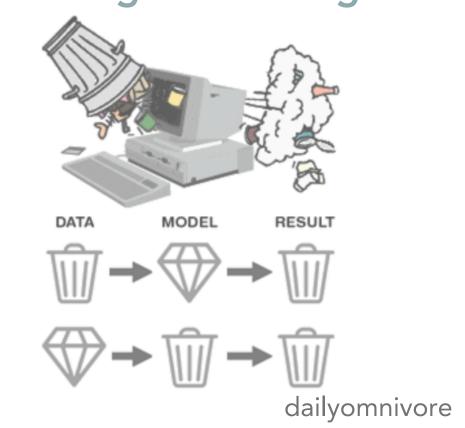
90

	count	unique	top	freq
Gender	520	2	Male	328
Polyuria	520	2	No	262
Polydipsia	520	2	No	287
sudden weight loss	520	2	No	303
weakness	520	2	Yes	305
Polyphagia	520	2	No	283
Genital thrush	520	2	No	404
visual blurring	520	2	No	287
Itching	520	2	No	267
Irritability	520	2	No	394
delayed healing	520	2	No	281
partial paresis	520	2	No	296
muscle stiffness	520	2	No	325
Alopecia	520	2	No	341
Obesity	520	2	No	432
class	520	2	Positive	320



PERSIAPAN DATA (DATA PREPARATION)

"Garbage in Garbage out"



Pemilihan Data

Menentukan dataset relevan



Integrasi & Konstruksi Data

- Mengabungkan data dari berbagai sumber
- Konstruksi data baru



Pembersihan Data

Melakukan koreksi pada:

- Data duplikat
- Data hilang
- Data keliru, typo
- Konsistensi Format



PERSIAPAN DATA (DATA PREPARATION)

Data cleaning Eliminasi data duplikat

```
# duplicates check
df.duplicated().sum()

269

# drop them
df = df.drop_duplicates()

# Sanity Check
df.duplicated().sum()
```

Data Preprocessing (data kategorik --> data numerik)

Perubahan format data

```
# Changing format
y = y.replace({'Positive':1, 'Negative': 0})
y.unique()
array([1, 0], dtype=int64)
```

PEMBUATAN MODEL MACHINE LEARNING



Beberapa model Machine Learning



Pemilihan Teknik Modelling

Mencari teknik modeling yg tepat untuk prediksi



Perancangan Pelatihan model

Merancang pelatihan model dan menentukan komposisi training & tes dataset



Pembangunan model

Membuat beberapa model & membandingkan performanya

Tentukan target & fitur, set proporsi pembagian data latihan dan test

```
# Define features and target
X = data.drop(columns ='class')
y = data['class']

# Train test split # 30% data test and 70% data train
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3, random_state=0, stratify=y)
```

Membangun model XGBoost & Random Forest

```
# Define model Random Forest
# Define model XGBoost
from xgboost import XGBClassifier
                                                              from sklearn.ensemble import RandomForestClassifier
xgboost = XGBClassifier()
                                                              rfc = RandomForestClassifier()
# Creating pipeline
                                                              # Fitting the modeL
                                                              from imblearn.pipeline import Pipeline
from imblearn.pipeline import Pipeline
classification_pipeline_xgb = Pipeline(
                                                              classification pipeline rfc = Pipeline(
   [('transformer', transformer),
                                                                  [('transformer', transformer),
    ('model', xgboost)
                                                                   ('model', rfc)
# Fitting model to XGBoost
                                                              # Fitting model to random forest
classification_pipeline_xgb.fit(X_train, y_train)
                                                              classification_pipeline_rfc.fit(X_train, y_train)
                     Pipeline
                                                                                   Pipeline
         transformer: ColumnTransformer
                                                                       transformer: ColumnTransformer
             onehot
                                 remainder
                                                                                                remainder
                                                                            onehot
                                                                                              ▶ passthrough

    OneHotEncoder

                               passthrough
                                                                      OneHotEncoder
                 ▶ XGBClassifier
                                                                         RandomForestClassifier
```

Evaluasi Model

- Bagaimana kualitas model?
- Apakah memenuhi standard & kriteria bisnis?

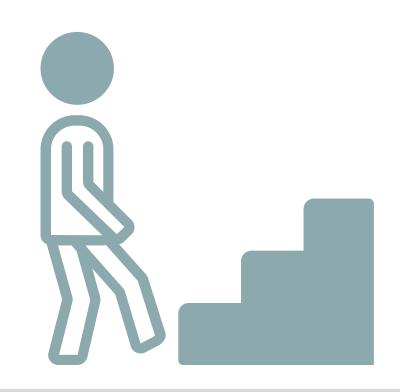
Review Proses

- Masih adakah kekurangan?
- Apakah semua proses tereksekusi dengan baik
- Buat summary dari temuan model dan koreksi

Tentukan langkah lanjutan

- Lanjutkan ke deployment?
- Perbaikan/Iterasi ulang model?





Evaluasi model

XGBoost

```
# XGB model performance
from sklearn.metrics import recall_score

# performance on training data
y_train_pred = classification_pipeline_xgb.predict(X_train)
print(f'Random Forest recall on train data: {recall_score(y_train,y_train_pred):.4f}')

# performance on testing data
y_test_pred = classification_pipeline_xgb.predict(X_test)
print(f'Random Forest recall on test data: {recall_score(y_test,y_test_pred):.4f}')

Random Forest recall on train data: 0.9917
Random Forest recall on test data: 0.9423
```

Random Forest

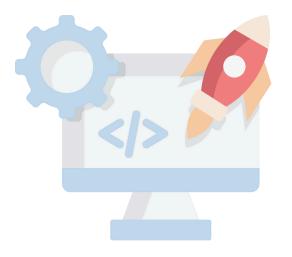
```
# Random forest model performance
from sklearn.metrics import recall_score

# performance on training data
y_train_pred = classification_pipeline_rfc.predict(X_train)
print(f'Random Forest recall on train data: {recall_score(y_train,y_train_pred):.4f}')

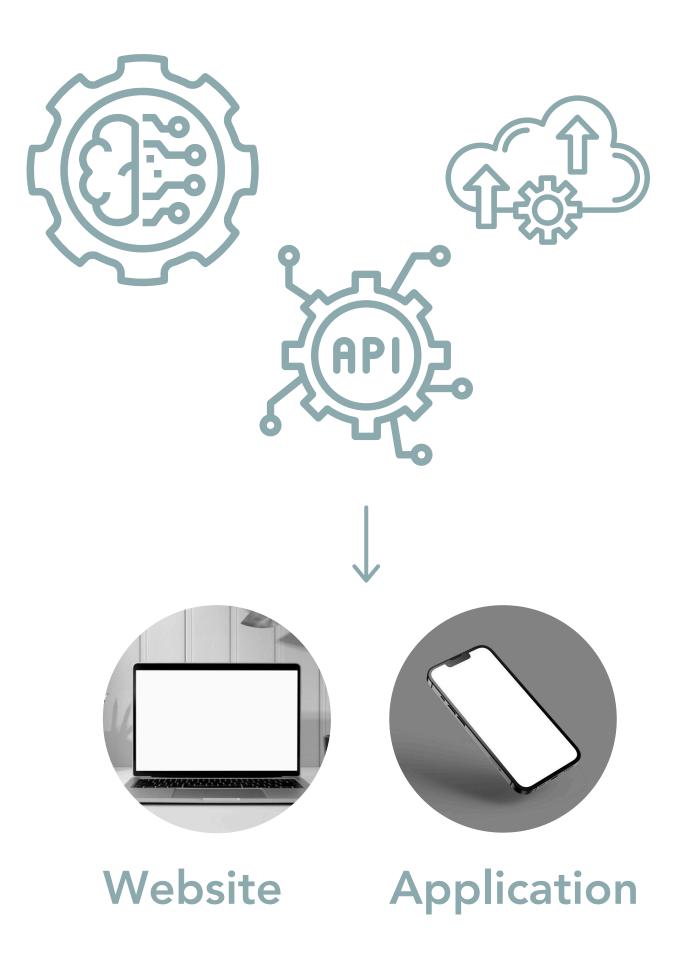
# performance on testing data
y_test_pred = classification_pipeline_rfc.predict(X_test)
print(f'Random Forest recall on test data: {recall_score(y_test,y_test_pred):.4f}')

Random Forest recall on train data: 1.0000
Random Forest recall on test data: 0.9808
```

MODEL DEPLOYMENT



- Perencanaan deployment
 - Model terpilih disimpan (format pickle)
 - Penggunaan layanan cloud
 - Penyajian API
 - Integrasi ke Apps atau Website
- Monitoring & Maintenance



Menyimpan model

```
import pickle

# Simpan model
pickle.dump(classification_pipeline_rfc, open("random_forest_diabetes_model.pkl", "wb"))
```

Membuat API untuk deployment

```
import pickle
import numpy as np
from flask import Flask, request, jsonify
# Load the trained model
model = pickle.load(open('random forest diabetes model.pkl', 'rb'))
# Initialize Flask app
app = Flask(__name__)
@app.route('/')
def home():
    return "Diabetes Prediction API is running!"
@app.route('/predict', methods=['POST'])
def predict():
    data = request.get json() # Expecting JSON input
    features = np.array(data['features']).reshape(1, -1) # Convert input to numpy array
    prediction = model.predict(features) # Make prediction
    return jsonify({'prediction': int(prediction[0])}) # Return prediction as JSON
if name == ' main ':
    app.run(debug=True) # Run API on Localhost
```

Aplikasi pendeteksi risiko diabetes

Diabetes risk predictor apps

This apps predicts early symptoms of diabetes melitus

The dataset for this prediction was obtained from Diabetes symptoms dataset by UC Irvine ML reposito





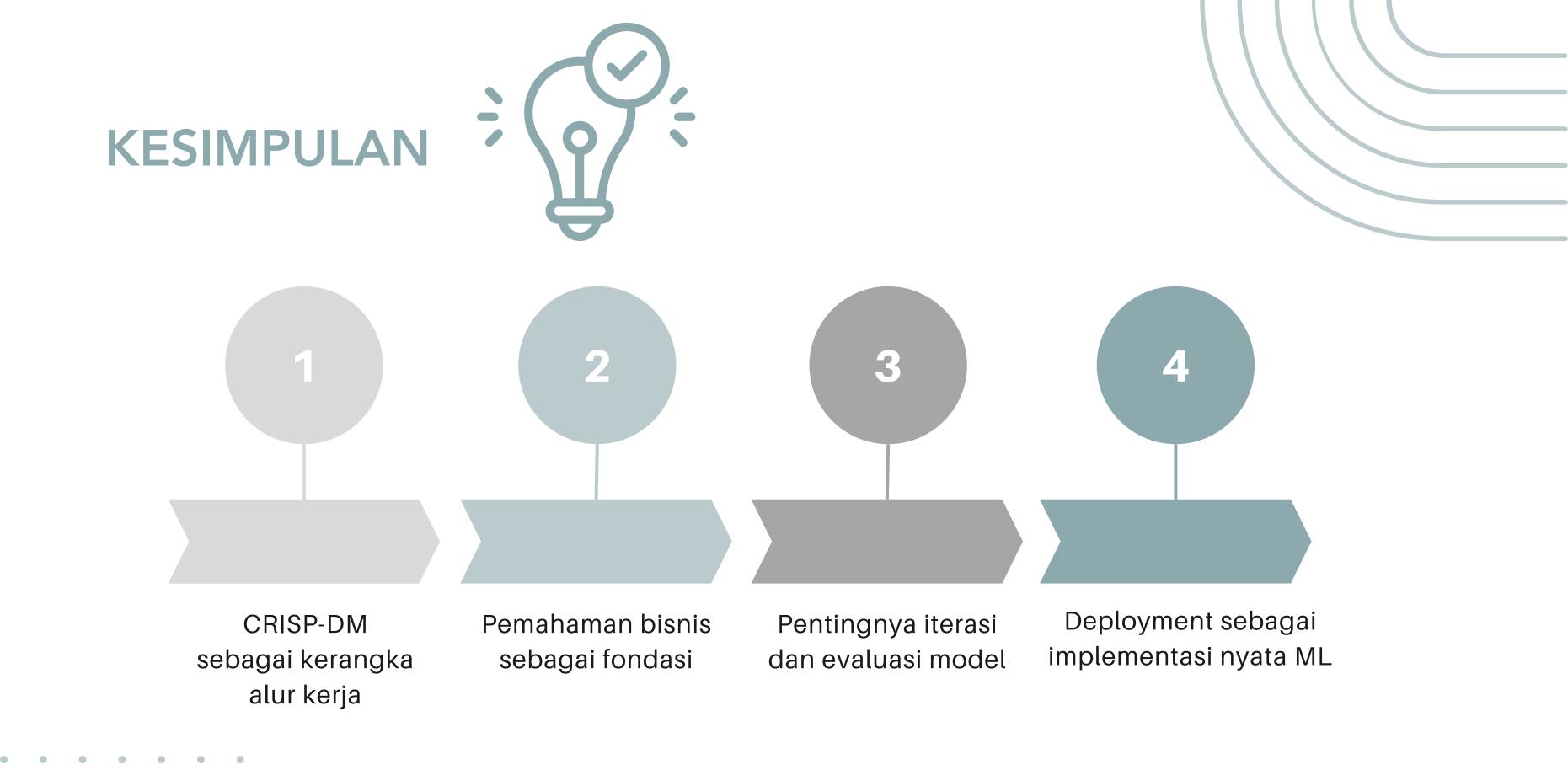
Prediksi Positif



7

Prediksi Negatif





THANKYOU

Do you have any question?

My Contact



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