```
import pandas as pd
from sklearn.linear model import LogisticRegression
from sklearn.semi supervised import SelfTrainingClassifier
from sklearn.preprocessing import LabelEncoder
import numpy as np
df=pd.read csv('/content/titanic.csv')
df = df[["Survived", "Pclass", "Sex", "Age", "Fare"]].dropna()
le=LabelEncoder()
df["Sex"]=le.fit transform(df["Sex"])
np.random.seed(42)
mask = np.random.rand(len(df)) < 0.5
df.loc[mask, "Survived"] = -1
X = df.drop("Survived", axis=1)
y = df["Survived"]
base model=LogisticRegression()
self training model=SelfTrainingClassifier(base model,criterion='k bes
t')
self training model.fit(X, y)
SelfTrainingClassifier(criterion='k best',
estimator=LogisticRegression())
df["predicted model"]=self training model.predict(X)
print(df.head(20))
   Survived Pclass Sex Age
                                          predicted model
                                    Fare
                   3
                       1
                          34.5
0
          - 1
                                  7.8292
                                                        0
1
           1
                   3
                        0 47.0
                                  7.0000
                                                        1
2
                   2
           0
                       1 62.0
                                  9.6875
                                                        0
3
           0
                   3
                       1 27.0 8.6625
                                                        0
4
          - 1
                   3
                        0 22.0 12.2875
                                                        1
5
                   3
          - 1
                       1 14.0
                                                        0
                                  9.2250
6
                   3
                        0 30.0
          - 1
                                  7.6292
                                                        1
7
                   2
           0
                       1 26.0 29.0000
                                                        0
                   3
8
          1
                        0 18.0
                                 7.2292
                                                        1
9
                   3
                       1 21.0 24.1500
           0
                                                        0
11
          - 1
                  1
                       1 46.0 26.0000
                                                        0
           1
                        0 23.0 82.2667
12
                   1
                                                        1
                   2
13
                       1 63.0 26.0000
                                                        0
           0
14
                   1
                        0 47.0 61.1750
                                                        1
          - 1
                   2
15
          - 1
                        0 24.0 27.7208
                                                        1
16
          - 1
                   2
                        1 35.0 12.3500
                                                        0
                   3
                                  7.2250
17
          - 1
                           21.0
                                                        0
```

18	1	3	0	27.0	7.9250	1
19	-1	3	0	45.0	7.2250	1
20	-1	1	1	55.0	59.4000	0