Crime Prediction Using Machine Learning

Crime prediction is an emerging field of study that uses data analysis and machine learning to predict criminal behavior and prevent crime.

Predictive models use advanced algorithms to identify patterns in criminal behavior and take into account external factors to make more accurate predictions.

Predictive models can help law enforcement agencies allocate resources, prevent crime, and improve public safety, as well as help investigators identify potential suspects and gather evidence more efficiently.

Researchers and practitioners must be aware of ethical and legal concerns surrounding crime prediction, such as potential biases and privacy and civil liberties, and address them through transparent and ethical practices.

Training Dataset Information

For more information click on the following link:

Dataset Info

Attribute Information:

Attribute Information: (122 predictive, 5 non-predictive, 1 goal)

-- state: US state (by number) - not counted as predictive above, but if considered, should be consided nominal (nominal) -- county: numeric code for county - not predictive, and many missing values (numeric) -- community: numeric code for community - not predictive and many missing values (numeric) -- communityname: community name - not predictive - for information only (string) -- fold: fold number for non-random 10 fold cross validation, potentially useful for debugging, paired tests - not predictive (numeric) -- population: population for community: (numeric - decimal) -- householdsize: mean people per household (numeric - decimal) -- racepctblack: percentage of population that is african american (numeric - decimal) -- racePctWhite: percentage of population that is caucasian (numeric - decimal) -- racePctAsian: percentage of population that is of asian heritage (numeric - decimal) -- racePctHisp: percentage of population that is of hispanic heritage (numeric - decimal) -- agePct12t21: percentage of population that is 12-21 in age (numeric - decimal) -- agePct12t29: percentage of population that is 12-29 in age (numeric - decimal) -- agePct16t24: percentage of population that is 16-24 in age (numeric - decimal) -- agePct65up: percentage of population that is 65 and over in age (numeric - decimal) -- numbUrban: number of people living in areas classified as urban (numeric - decimal)

- -- pctUrban: percentage of people living in areas classified as urban (numeric decimal)
- -- medIncome: median household income (numeric decimal)
- -- pctWWage: percentage of households with wage or salary income in 1989 (numeric decimal)
- -- pctWFarmSelf: percentage of households with farm or self employment income in 1989 (numeric decimal)
- -- pctWInvInc: percentage of households with investment / rent income in 1989 (numeric decimal)
- -- pctWSocSec: percentage of households with social security income in 1989 (numeric decimal)
- -- pctWPubAsst: percentage of households with public assistance income in 1989 (numeric decimal)
- -- pctWRetire: percentage of households with retirement income in 1989 (numeric decimal)
- -- medFamInc: median family income (differs from household income for non-family households) (numeric decimal)
- -- perCapInc: per capita income (numeric decimal)
- -- whitePerCap: per capita income for caucasians (numeric decimal)
- -- blackPerCap: per capita income for african americans (numeric decimal)
- -- indianPerCap: per capita income for native americans (numeric decimal)
- -- AsianPerCap: per capita income for people with asian heritage (numeric decimal)
- -- OtherPerCap: per capita income for people with 'other' heritage (numeric decimal)
- -- HispPerCap: per capita income for people with hispanic heritage (numeric decimal)

- -- NumUnderPov: number of people under the poverty level (numeric decimal)
- -- PctPopUnderPov: percentage of people under the poverty level (numeric decimal)
- -- PctLess9thGrade: percentage of people 25 and over with less than a 9th grade education (numeric decimal)
- -- PctNotHSGrad: percentage of people 25 and over that are not high school graduates (numeric decimal)
- -- PctBSorMore: percentage of people 25 and over with a bachelors degree or higher education (numeric decimal)
- -- PctUnemployed: percentage of people 16 and over, in the labor force, and unemployed (numeric decimal)
- -- PctEmploy: percentage of people 16 and over who are employed (numeric decimal)
- -- PctEmplManu: percentage of people 16 and over who are employed in manufacturing (numeric decimal)
- -- PctEmplProfServ: percentage of people 16 and over who are employed in professional services (numeric decimal)
- -- PctOccupManu: percentage of people 16 and over who are employed in manufacturing (numeric decimal) #######
- -- PctOccupMgmtProf: percentage of people 16 and over who are employed in management or professional occupations (numeric decimal)
- -- MalePctDivorce: percentage of males who are divorced (numeric decimal)
- -- MalePctNevMarr: percentage of males who have never married (numeric decimal)

- -- FemalePctDiv: percentage of females who are divorced (numeric decimal)
- -- TotalPctDiv: percentage of population who are divorced (numeric decimal)
- -- PersPerFam: mean number of people per family (numeric decimal)
- -- PctFam2Par: percentage of families (with kids) that are headed by two parents (numeric decimal)
- -- PctKids2Par: percentage of kids in family housing with two parents (numeric decimal)
- -- PctYoungKids2Par: percent of kids 4 and under in two parent households (numeric decimal)
- -- PctTeen2Par: percent of kids age 12-17 in two parent households (numeric decimal)
- -- PctWorkMomYoungKids: percentage of moms of kids 6 and under in labor force (numeric decimal)
- -- PctWorkMom: percentage of moms of kids under 18 in labor force (numeric decimal)
- -- NumIlleg: number of kids born to never married (numeric decimal)
- -- Pctllleg: percentage of kids born to never married (numeric decimal)
- -- NumImmig: total number of people known to be foreign born (numeric decimal)
- -- PctImmigRecent: percentage of immigrants who immigated within last 3 years (numeric decimal)
- -- PctImmigRec5: percentage of immigrants who immigated within last 5 years (numeric decimal)
- -- PctImmigRec8: percentage of immigrants who immigated within last 8 years (numeric decimal)
- -- PctImmigRec10: percentage of *immigrants* who immigated within last 10 years (numeric decimal)
- -- PctRecentImmig: percent of population who have immigrated within the last 3 years (numeric decimal)

- -- PctRecImmig5: percent of *population* who have immigrated within the last 5 years (numeric decimal)
- -- PctRecImmig8: percent of *population* who have immigrated within the last 8 years (numeric decimal)
- -- PctRecImmig10: percent of *population* who have immigrated within the last 10 years (numeric decimal)
- -- PctSpeakEnglOnly: percent of people who speak only English (numeric decimal)
- -- PctNotSpeakEnglWell: percent of people who do not speak English well (numeric decimal)
- -- PctLargHouseFam: percent of family households that are large (6 or more) (numeric decimal)
- -- PctLargHouseOccup: percent of all occupied households that are large (6 or more people) (numeric decimal)
- -- PersPerOccupHous: mean persons per household (numeric decimal)
- -- PersPerOwnOccHous: mean persons per owner occupied household (numeric decimal)
- -- PersPerRentOccHous: mean persons per rental household (numeric decimal)
- -- PctPersOwnOccup: percent of people in owner occupied households (numeric decimal)
- -- PctPersDenseHous: percent of persons in dense housing (more than 1 person per room) (numeric decimal)
- -- PctHousLess3BR: percent of housing units with less than 3 bedrooms (numeric decimal)
- -- MedNumBR: median number of bedrooms (numeric decimal)
- -- Hous Vacant: number of vacant households (numeric decimal)

- -- PctHousOccup: percent of housing occupied (numeric decimal)
- -- PctHousOwnOcc: percent of households owner occupied (numeric decimal)
- -- PctVacantBoarded: percent of vacant housing that is boarded up (numeric decimal)
- -- PctVacMore6Mos: percent of vacant housing that has been vacant more than 6 months (numeric decimal)
- -- MedYrHousBuilt: median year housing units built (numeric decimal)
- -- PctHousNoPhone: percent of occupied housing units without phone (in 1990, this was rare!) (numeric decimal)
- -- PctWOFullPlumb: percent of housing without complete plumbing facilities (numeric decimal)
- -- OwnOccLowQuart: owner occupied housing lower quartile value (numeric decimal)
- -- OwnOccMedVal: owner occupied housing median value (numeric decimal)
- -- OwnOccHiQuart: owner occupied housing upper quartile value (numeric decimal)
- -- RentLowQ: rental housing lower quartile rent (numeric decimal)
- -- RentMedian: rental housing median rent (Census variable H32B from file STF1A) (numeric decimal)
- -- RentHighQ: rental housing upper quartile rent (numeric decimal)
- -- MedRent: median gross rent (Census variable H43A from file STF3A includes utilities) (numeric decimal)
- -- MedRentPctHousInc: median gross rent as a percentage of household income (numeric decimal)

- -- MedOwnCostPctInc: median owners cost as a percentage of household income for owners with a mortgage (numeric decimal)
- -- MedOwnCostPctIncNoMtg: median owners cost as a percentage of household income for owners without a mortgage (numeric decimal)
- -- NumInShelters: number of people in homeless shelters (numeric decimal)
- -- NumStreet: number of homeless people counted in the street (numeric decimal)
- -- PctForeignBorn: percent of people foreign born (numeric decimal)
- -- PctBornSameState: percent of people born in the same state as currently living (numeric decimal)
- -- PctSameHouse85: percent of people living in the same house as in 1985 (5 years before) (numeric decimal)
- -- PctSameCity85: percent of people living in the same city as in 1985 (5 years before) (numeric decimal)
- -- PctSameState85: percent of people living in the same state as in 1985 (5 years before) (numeric decimal)
- -- LemasSwornFT: number of sworn full time police officers (numeric decimal)
- -- LemasSwFTPerPop: sworn full time police officers per 100K population (numeric decimal)
- -- LemasSwFTFieldOps: number of sworn full time police officers in field operations (on the street as opposed to administrative etc) (numeric decimal)
- -- LemasSwFTFieldPerPop: sworn full time police officers in field operations (on the street as opposed to administrative etc) per 100K population (numeric decimal)

- -- LemasTotalReq: total requests for police (numeric decimal)
- -- LemasTotReqPerPop: total requests for police per 100K popuation (numeric decimal)
- -- PolicReqPerOffic: total requests for police per police officer (numeric decimal)
- -- PolicPerPop: police officers per 100K population (numeric decimal)
- -- RacialMatchCommPol: a measure of the racial match between the community and the police force. High values indicate proportions in community and police force are similar (numeric decimal)
- -- PctPolicWhite: percent of police that are caucasian (numeric decimal)
- -- PctPolicBlack: percent of police that are african american (numeric decimal)
- -- PctPolicHisp: percent of police that are hispanic (numeric decimal)
- -- PctPolicAsian: percent of police that are asian (numeric decimal)
- -- PctPolicMinor: percent of police that are minority of any kind (numeric decimal)
- -- OfficAssgnDrugUnits: number of officers assigned to special drug units (numeric decimal)
- -- NumKindsDrugsSeiz: number of different kinds of drugs seized (numeric decimal)
- -- PolicAveOTWorked: police average overtime worked (numeric decimal)
- -- LandArea: land area in square miles (numeric decimal)
- -- PopDens: population density in persons per square mile (numeric decimal)
- -- PctUsePubTrans: percent of people using public transit for commuting (numeric decimal)

- -- PolicCars: number of police cars (numeric decimal)
- -- PolicOperBudg: police operating budget (numeric decimal)
- -- LemasPctPolicOnPatr: percent of sworn full time police officers on patrol (numeric decimal)
- -- LemasGangUnitDeploy: gang unit deployed (numeric decimal but really ordinal 0 means NO, 1 means YES, 0.5 means Part Time)
- -- LemasPctOfficDrugUn: percent of officers assigned to drug units (numeric decimal)
- -- PolicBudgPerPop: police operating budget per population (numeric decimal)
- -- ViolentCrimesPerPop: total number of violent crimes per 100K popuation (numeric decimal) GOAL attribute (to be predicted)

Summary Statistics:

Min Max Mean SD Correl Median Mode Missing

population 0 1 0.06 0.13 0.37 0.02 0.01 0

householdsize 0 1 0.46 0.16 -0.03 0.44 0.41 0

racepctblack 0 1 0.18 0.25 0.63 0.06 0.01 0

racePctWhite 0 1 0.75 0.24 -0.68 0.85 0.98 0

racePctAsian 0 1 0.15 0.21 0.04 0.07 0.02 0

racePctHisp 0 1 0.14 0.23 0.29 0.04 0.01 0

agePct12t21 0 1 0.42 0.16 0.06 0.4 0.38 0

agePct12t29 0 1 0.49 0.14 0.15 0.48 0.49 0 agePct16t24 0 1 0.34 0.17 0.10 0.29 0.29 0 agePct65up 0 1 0.42 0.18 0.07 0.42 0.47 0 numbUrban 0 1 0.06 0.13 0.36 0.03 0 0 pctUrban 0 1 0.70 0.44 0.08 1 1 0 medIncome 0 1 0.36 0.21 -0.42 0.32 0.23 0 pctWWage 0 1 0.56 0.18 -0.31 0.56 0.58 0 pctWFarmSelf 0 1 0.29 0.20 -0.15 0.23 0.16 0 pctWInvInc 0 1 0.50 0.18 -0.58 0.48 0.41 0 pctWSocSec 0 1 0.47 0.17 0.12 0.475 0.56 0 pctWPubAsst 0 1 0.32 0.22 0.57 0.26 0.1 0 pctWRetire 0 1 0.48 0.17 -0.10 0.47 0.44 0 medFamInc 0 1 0.38 0.20 -0.44 0.33 0.25 0 perCapInc 0 1 0.35 0.19 -0.35 0.3 0.23 0 whitePerCap 0 1 0.37 0.19 -0.21 0.32 0.3 0 blackPerCap 0 1 0.29 0.17 -0.28 0.25 0.18 0 indianPerCap 0 1 0.20 0.16 -0.09 0.17 0 0

AsianPerCap 0 1 0.32 0.20 -0.16 0.28 0.18 0 OtherPerCap 0 1 0.28 0.19 -0.13 0.25 0 1 HispPerCap 0 1 0.39 0.18 -0.24 0.345 0.3 0 NumUnderPov 0 1 0.06 0.13 0.45 0.02 0.01 0 PctPopUnderPov 0 1 0.30 0.23 0.52 0.25 0.08 0 PctLess9thGrade 0 1 0.32 0.21 0.41 0.27 0.19 0 PctNotHSGrad 0 1 0.38 0.20 0.48 0.36 0.39 0 PctBSorMore 0 1 0.36 0.21 -0.31 0.31 0.18 0 PctUnemployed 0 1 0.36 0.20 0.50 0.32 0.24 0 PctEmploy 0 1 0.50 0.17 -0.33 0.51 0.56 0 PctEmplManu 0 1 0.40 0.20 -0.04 0.37 0.26 0 PctEmplProfServ 0 1 0.44 0.18 -0.07 0.41 0.36 0 PctOccupManu 0 1 0.39 0.20 0.30 0.37 0.32 0 PctOccupMgmtProf 0 1 0.44 0.19 -0.34 0.4 0.36 0 MalePctDivorce 0 1 0.46 0.18 0.53 0.47 0.56 0 MalePctNevMarr 0 1 0.43 0.18 0.30 0.4 0.38 0 FemalePctDiv 0 1 0.49 0.18 0.56 0.5 0.54 0

TotalPctDiv 0 1 0.49 0.18 0.55 0.5 0.57 0

PersPerFam 0 1 0.49 0.15 0.14 0.47 0.44 0

PctFam2Par 0 1 0.61 0.20 -0.71 0.63 0.7 0

PctKids2Par 0 1 0.62 0.21 -0.74 0.64 0.72 0

PctYoungKids2Par 0 1 0.66 0.22 -0.67 0.7 0.91 0

PctTeen2Par 0 1 0.58 0.19 -0.66 0.61 0.6 0

PctWorkMomYoungKids 0 1 0.50 0.17 -0.02 0.51 0.51 0

PctWorkMom 0 1 0.53 0.18 -0.15 0.54 0.57 0

NumIlleg 0 1 0.04 0.11 0.47 0.01 0 0

Pctllleg 0 1 0.25 0.23 0.74 0.17 0.09 0

Numlmmig 0 1 0.03 0.09 0.29 0.01 0 0

PctImmigRecent 0 1 0.32 0.22 0.17 0.29 0 0

PctlmmigRec5 0 1 0.36 0.21 0.22 0.34 0 0

PctImmigRec8 0 1 0.40 0.20 0.25 0.39 0.26 0

PctlmmigRec10 0 1 0.43 0.19 0.29 0.43 0.43 0

PctRecentImmig 0 1 0.18 0.24 0.23 0.09 0.01 0

PctRecImmig5 0 1 0.18 0.24 0.25 0.08 0.02 0

PctRecImmig8 0 1 0.18 0.24 0.25 0.09 0.02 0 PctRecImmig10 0 1 0.18 0.23 0.26 0.09 0.02 0 PctSpeakEnglOnly 0 1 0.79 0.23 -0.24 0.87 0.96 0 PctNotSpeakEnglWell 0 1 0.15 0.22 0.30 0.06 0.03 0 PctLargHouseFam 0 1 0.27 0.20 0.38 0.2 0.17 0 PctLargHouseOccup 0 1 0.25 0.19 0.29 0.19 0.19 0 PersPerOccupHous 0 1 0.46 0.17 -0.04 0.44 0.37 0 PersPerOwnOccHous 0 1 0.49 0.16 -0.12 0.48 0.45 0 PersPerRentOccHous 0 1 0.40 0.19 0.25 0.36 0.32 0 PctPersOwnOccup 0 1 0.56 0.20 -0.53 0.56 0.54 0 PctPersDenseHous 0 1 0.19 0.21 0.45 0.11 0.06 0 PctHousLess3BR 0 1 0.50 0.17 0.47 0.51 0.53 0 MedNumBR 0 1 0.31 0.26 -0.36 0.5 0.5 0 HousVacant 0 1 0.08 0.15 0.42 0.03 0.01 0 PctHousOccup 0 1 0.72 0.19 -0.32 0.77 0.88 0 PctHousOwnOcc 0 1 0.55 0.19 -0.47 0.54 0.52 0 PctVacantBoarded 0 1 0.20 0.22 0.48 0.13 0 0

PctVacMore6Mos 0 1 0.43 0.19 0.02 0.42 0.44 0

MedYrHousBuilt 0 1 0.49 0.23 -0.11 0.52 0 0

PctHousNoPhone 0 1 0.26 0.24 0.49 0.185 0.01 0

PctWOFullPlumb 0 1 0.24 0.21 0.36 0.19 0 0

OwnOccLowQuart 0 1 0.26 0.22 -0.21 0.18 0.09 0

OwnOccMedVal 0 1 0.26 0.23 -0.19 0.17 0.08 0

OwnOccHiQuart 0 1 0.27 0.24 -0.17 0.18 0.08 0

RentLowQ 0 1 0.35 0.22 -0.25 0.31 0.13 0

RentMedian 0 1 0.37 0.21 -0.24 0.33 0.19 0

RentHighQ 0 1 0.42 0.25 -0.23 0.37 1 0

MedRent 0 1 0.38 0.21 -0.24 0.34 0.17 0

MedRentPctHousInc 0 1 0.49 0.17 0.33 0.48 0.4 0

MedOwnCostPctInc 0 1 0.45 0.19 0.06 0.45 0.41 0

MedOwnCostPctIncNoMtg 0 1 0.40 0.19 0.05 0.37 0.24 0

NumInShelters 0 1 0.03 0.10 0.38 0 0 0

NumStreet 0 1 0.02 0.10 0.34 0 0 0

PctForeignBorn 0 1 0.22 0.23 0.19 0.13 0.03 0

PctBornSameState 0 1 0.61 0.20 -0.08 0.63 0.78 0

PctSameHouse85 0 1 0.54 0.18 -0.16 0.54 0.59 0

PctSameCity85 0 1 0.63 0.20 0.08 0.67 0.74 0

PctSameState85 0 1 0.65 0.20 -0.02 0.7 0.79 0

LemasSwornFT 0 1 0.07 0.14 0.34 0.02 0.02 1675

LemasSwFTPerPop 0 1 0.22 0.16 0.15 0.18 0.2 1675

LemasSwFTFieldOps 0 1 0.92 0.13 -0.33 0.97 0.98 1675

LemasSwFTFieldPerPop 0 1 0.25 0.16 0.16 0.21 0.19 1675

LemasTotalReq 0 1 0.10 0.16 0.35 0.04 0.02 1675

LemasTotReqPerPop 0 1 0.22 0.16 0.27 0.17 0.14 1675

PolicReqPerOffic 0 1 0.34 0.20 0.17 0.29 0.23 1675

PolicPerPop 0 1 0.22 0.16 0.15 0.18 0.2 1675

RacialMatchCommPol 0 1 0.69 0.23 -0.46 0.74 0.78 1675

PctPolicWhite 0 1 0.73 0.22 -0.44 0.78 0.72 1675

PctPolicBlack 0 1 0.22 0.24 0.54 0.12 0 1675

PctPolicHisp 0 1 0.13 0.20 0.12 0.06 0 1675

PctPolicAsian 0 1 0.11 0.23 0.10 0 0 1675

PctPolicMinor 0 1 0.26 0.23 0.49 0.2 0.07 1675

OfficAssgnDrugUnits 0 1 0.08 0.12 0.34 0.04 0.03 1675

NumKindsDrugsSeiz 0 1 0.56 0.20 0.13 0.57 0.57 1675

PolicAveOTWorked 0 1 0.31 0.23 0.03 0.26 0.19 1675

LandArea 0 1 0.07 0.11 0.20 0.04 0.01 0

PopDens 0 1 0.23 0.20 0.28 0.17 0.09 0

PctUsePubTrans 0 1 0.16 0.23 0.15 0.07 0.01 0

PolicCars 0 1 0.16 0.21 0.38 0.08 0.02 1675

PolicOperBudg 0 1 0.08 0.14 0.34 0.03 0.02 1675

LemasPctPolicOnPatr 0 1 0.70 0.21 -0.08 0.75 0.74 1675

LemasGangUnitDeploy 0 1 0.44 0.41 0.12 0.5 0 1675

ML model used: Randam Forest

PolicBudgPerPop 0_1 0.20 0.16 0.10 0.15 0.12 1675

Algorithm
ViolenterimesPerPop 0 1 0.24 0.23 1.00 0.15 0.03 0

Random Forest is a popular machine learning algorithm that is used for both classification and regression tasks. It is an ensemble method that combines multiple decision trees to make predictions. The idea behind the algorithm is to generate a large number of decision trees, where each tree is trained on a randomly sampled subset of the training data and a random subset of the input features.

During prediction, each decision tree in the forest independently makes a prediction, and the final prediction is then determined by taking the majority vote of all the trees. This helps to reduce overfitting and improve the accuracy of the predictions.

Random Forest is a versatile algorithm that can be used for a wide range of applications, including image classification, fraud detection, and customer churn prediction. It is also relatively easy to use, as it requires minimal data preprocessing and hyperparameter tuning. Overall, Random Forest is a powerful and popular algorithm in the field of machine learning.

Enter the 100 features according to given info provided in the dataset section in the form (feature values seprated by ,)

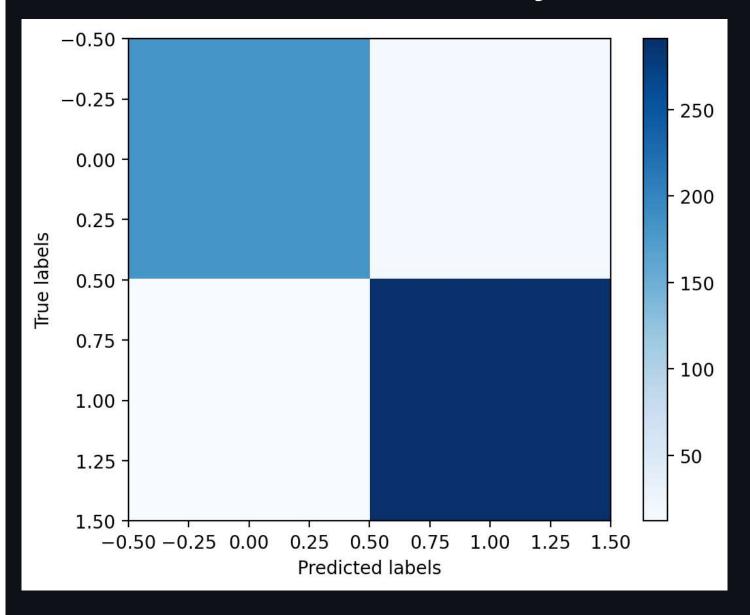
Enter Input Array

0.05, 0.67, 0.04, 0.64, 0.5, 0.46, 0.52, 0.61, 0.45, 0.2, 0.07, 1, 0.65, 0.82, 0.22, 0.56, 0.2, 0.2, 0.35, 0.62, 0.47, 0.53, 0.37, 0.20,

Submit

High Chance of Crime

Confusin Matrix and accuracy score



Accuracy of model is : 0.944 or 94.389%

This confusion matrix represents the performance of a binary classification model on a test dataset. The rows of the matrix correspond to the actual class labels, while the columns correspond to the predicted class labels.

In this particular case, the model predicted 196 instances to be positive (belonging to the second class) and 303 instances to be negative (belonging to the first class). Out of the 196 positive instances, the model correctly predicted 180, but misclassified 16 as negative. Out of the 303 negative instances, the model correctly predicted 291, but misclassified 12 as positive.

Therefore, the confusion matrix shows that the model has a high true positive rate (TPR) of 0.938 and a high true negative rate (TNR) of 0.960, but a relatively low precision (0.948) and F1-score (0.943) due to the false positive and false negative classifications.

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