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"3 15701354 Boni 699 France Female 39 1 \\n",
"4 15737888 Mitchell 850 Spain Female 43 2 \\n",
"\\n",
" Balance Num Of Products Has Credit Card Is Active Member \\n",
"0 0.00 1 1 1 \\n",
"1 83807.86 1 0 1 \\n",
"2 159660.80 3 1 0 \\n",
"3 0.00 2 0 0 \\n",
"4 125510.82 1 1 1 \\n",
"\\n",
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"0 101348.88 1 \\n",
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        "Data columns (total 13 columns):\n",
        "#   Column          Non-Null Count  Dtype  \n",
        "---  -

```

#	Column	Non-Null Count	Dtype
0	CustomerId	10000 non-null	int64
1	Surname	10000 non-null	object
2	CreditScore	10000 non-null	int64
3	Geography	10000 non-null	object
4	Gender	10000 non-null	object
5	Age	10000 non-null	int64
6	Tenure	10000 non-null	int64
7	Balance	10000 non-null	float64
8	Num Of Products	10000 non-null	int64
9	Has Credit Card	10000 non-null	int64
10	Is Active Member	10000 non-null	int64
11	Estimated Salary	10000 non-null	float64
12	Churn	10000 non-null	int64

```

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        "---  -
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        1   CreditScore      10000 non-null int64  \n",
        2   Geography        10000 non-null object \n",
        3   Gender           10000 non-null object \n",
        4   Age              10000 non-null int64  \n",
        5   Tenure           10000 non-null int64  \n",
        6   Balance          10000 non-null float64\n",
        7   Num Of Products  10000 non-null int64  \n",

```

```

" 8  Has Credit Card  10000 non-null int64 \n",
" 9  Is Active Member 10000 non-null int64 \n",
" 10 Estimated Salary 10000 non-null float64\n",
" 11 Churn            10000 non-null int64 \n",
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"Germany   2509\n",
"Spain     2477\n",
"Name: Geography, dtype: int64"
]
}
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}
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                    "Female  4543\n",
                    "Name: Gender, dtype: int64"
                ]
            },
            "execution_count": 13,
            "metadata": {},
            "output_type": "execute_result"
        }
    ],
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    ]
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    "id": "aea29223",
    "metadata": {},
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            "data": {

```

```

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      "2   4590\n",
      "3    266\n",
      "4     60\n",
      "Name: Num Of Products, dtype: int64"
    ]
  },
  "execution_count": 15,
  "metadata": {},
  "output_type": "execute_result"
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          "0   2945\n",
          "Name: Has Credit Card, dtype: int64"
        ]
      },
      "execution_count": 17,
      "metadata": {},
      "output_type": "execute_result"
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  "source": [
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  ]
}

```

```

]
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  "metadata": {},
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      "data": {
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          "Name: Is Active Member, dtype: int64"
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      "execution_count": 18,
      "metadata": {},
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  ],
  "source": [
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  ]
},
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      "execution_count": 19,
      "metadata": {},
      "output_type": "execute_result"
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```

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5o0Fs13AiJYA9w1NbwGOGB6QZDWwuk1+N8mdu7G9A4G/243lX5CcM+4t/tCs7O8sc5/nh
3m3zzInt/b5x6ebMVfColNVrQXWzsS6kmyqqomZWNdcMN/2F9zn+cJ9njlz5TTUVuDgoemlr
U+SNAZzJSyUB5YnOSTJ3sDJwLZZrkmS5o05cRqqqrYn+bfAVcBewLqqquq3HTc7l6aw5ZL7t
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WktuT3JLk6iTT3nM9V4z6mpgkv5ikkSz52yxH2eckJ7U/69uSfG7cNc60Ef7dfm2SryW5qf37ffx
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6rqG8C25xiyAriwBq4B9k9y0O5sc76Gxa5eH7JkujFVtR14DHj1WKrrxyj7PGwVg/8zmcs697k
dnh9cVV8eZ2E9GuXP+SeBn0zyf5Nck+TYsVXXj1H2+ePAh5NsAa4Efn08pc2a5/vfe6c58Zy
FxivJh4EJ4OdnU5Y+JXkJ8Cngl2e5lHFbwOBU1LsYHD1+l8mbq+rR2SyqZx8C/riqfjfJO4CLkr
ypqv5xtgubK+brkcUorw/54ZgkCxcgcu48lur6MdlrU5K8F/jPwAIV9eSYautL1z6/EngT8PUk9zA
4t7thj/kHuXPeQuwoap+UFV3A/+PQXjMVaPs8yrgUoCq+mtgHwYvGXyxmvFXJM3XsBjl9SE
bgJWt/QHhg9WuHM1Rnfuc5G3AZxkExVw/jw0d+1xVj1XVgVW1rKqWMbhOc0JVbZqdcmfE
KP9uf4nBUQVJDmRwWuquMdY400bZ528DRwEk+SkgYTE11irHawNwSrsr6kjgsaq6f3dW
OC9PQ9U0rw9Jciawqao2ABcwOFSdZHAh6eTZq3j3bjP/w34MeCydi3/21V1wqwVvZtG3Oc
XIRH3+Srg6CS3A08D/6Gq5uxR84j7/NvAHyb5TQYXu395Lv/PX5LPMwj8A9t1mDXASwGq6
g8YXJc5HpgEngBO3e1tzuF/XpKkMZmvp6EkSc+DYSFJ6mRYSJI6GRaSpE6GhSSpk2EhS
epkWEiSOv1/euBUvZREvkgAAAAASUVORK5CYII=\n",

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  "<Figure size 432x288 with 1 Axes>"
],
"metadata": {
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"output_type": "display_data"
},
"source": [
  "y_ros.plot(kind = 'hist')"
],
{
  "cell_type": "markdown",
  "id": "28a4438d",
  "metadata": {},
  "source": [
    "# Train Test Split"
  ]
},
{
  "cell_type": "code",
  "execution_count": 44,
  "id": "8ab9c13f",
  "metadata": {},
  "outputs": [],
  "source": [
    "from sklearn.model_selection import train_test_split"
  ]
},
```

```

{
  "cell_type": "markdown",
  "id": "7bb0a2f3",
  "metadata": {},
  "source": [
    "# Split Original Data"
  ]
},
{
  "cell_type": "code",
  "execution_count": 45,
  "id": "ed463468",
  "metadata": {},
  "outputs": [],
  "source": [
    "x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.3, random_state = 2529)"
  ]
},
{
  "cell_type": "markdown",
  "id": "9626e83f",
  "metadata": {},
  "source": [
    "# Split Random Under Sample Data"
  ]
},
{
  "cell_type": "code",
  "execution_count": 46,
  "id": "592a083c",
  "metadata": {},
  "outputs": [],
  "source": [
    "x_train_rus, x_test_rus, y_train_rus, y_test_rus = train_test_split(x_rus, y_rus, test_size = 0.3, random_state = 2529)"
  ]
},
{
  "cell_type": "markdown",
  "id": "f1473ec7",
  "metadata": {},
  "source": [
    "# Split Random Over Sample Data"
  ]
},
{
  "cell_type": "code",

```

```

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    "id": "5c232f8a",
    "metadata": {},
    "outputs": [],
    "source": [
        "x_train_ros, x_test_ros, y_train_ros, y_test_ros = train_test_split(x_ros, y_ros, test_size =
0.3, random_state = 2529)"
    ]
},
{
    "cell_type": "markdown",
    "id": "9a7f692b",
    "metadata": {},
    "source": [
        "# Standardize Features"
    ]
},
{
    "cell_type": "code",
    "execution_count": 48,
    "id": "3c366c7a",
    "metadata": {},
    "outputs": [],
    "source": [
        "from sklearn.preprocessing import StandardScaler"
    ]
},
{
    "cell_type": "code",
    "execution_count": 49,
    "id": "e1aeb20d",
    "metadata": {},
    "outputs": [],
    "source": [
        "sc = StandardScaler()"
    ]
},
{
    "cell_type": "markdown",
    "id": "7b3a51a6",
    "metadata": {},
    "source": [
        "# Standardize Original Data"
    ]
},
{
    "cell_type": "code",
    "execution_count": 50,

```

```

    "id": "c6f2e144",
    "metadata": {},
    "outputs": [],
    "source": [
        "x_train[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']] =
sc.fit_transform(x_train[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']])"
    ]
},
{
    "cell_type": "code",
    "execution_count": 51,
    "id": "f0b29530",
    "metadata": {},
    "outputs": [],
    "source": [
        "x_test[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']] =
sc.fit_transform(x_test[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']])"
    ]
},
{
    "cell_type": "markdown",
    "id": "bb736d39",
    "metadata": {},
    "source": [
        "# Standardize Random Under Sample Data"
    ]
},
{
    "cell_type": "code",
    "execution_count": 52,
    "id": "ddd5551c",
    "metadata": {},
    "outputs": [],
    "source": [
        "x_train_rus[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']] =
sc.fit_transform(x_train_rus[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']])"
    ]
},
{
    "cell_type": "code",
    "execution_count": 53,
    "id": "67a368f6",
    "metadata": {},
    "outputs": [],
    "source": [
        "x_test_rus[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']] =
sc.fit_transform(x_test_rus[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']])"
    ]
}

```

```

},
{
  "cell_type": "markdown",
  "id": "75d603df",
  "metadata": {},
  "source": [
    "# Standardize Random Over Sample Data"
  ]
},
{
  "cell_type": "code",
  "execution_count": 54,
  "id": "bc0893d8",
  "metadata": {},
  "outputs": [],
  "source": [
    "x_train_ros[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']] =
sc.fit_transform(x_train_ros[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']])"
  ]
},
{
  "cell_type": "code",
  "execution_count": 55,
  "id": "235d469c",
  "metadata": {},
  "outputs": [],
  "source": [
    "x_test_ros[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']] =
sc.fit_transform(x_test_ros[['CreditScore', 'Age', 'Tenure', 'Balance', 'Estimated Salary']])"
  ]
},
{
  "cell_type": "markdown",
  "id": "268b82f0",
  "metadata": {},
  "source": [
    "# Support vector machine Classifier"
  ]
},
{
  "cell_type": "code",
  "execution_count": 56,
  "id": "35830bc2",
  "metadata": {},
  "outputs": [],
  "source": [
    "from sklearn.svm import SVC"
  ]
}

```

```

},
{
  "cell_type": "code",
  "execution_count": 57,
  "id": "d439e53b",
  "metadata": {},
  "outputs": [],
  "source": [
    "svc = SVC()"
  ]
},
{
  "cell_type": "code",
  "execution_count": 58,
  "id": "4b4b329f",
  "metadata": {},
  "outputs": [
    {
      "data": {
        "text/plain": [
          "SVC()"
        ]
      },
      "execution_count": 58,
      "metadata": {},
      "output_type": "execute_result"
    }
  ],
  "source": [
    "svc.fit(x_train, y_train)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 59,
  "id": "194bb8cf",
  "metadata": {},
  "outputs": [],
  "source": [
    "y_pred = svc.predict(x_test)"
  ]
},
{
  "cell_type": "markdown",
  "id": "ce8cfdc5",
  "metadata": {},
  "source": [
    "# Model Accuracy"
  ]
}

```

```

]
},
{
  "cell_type": "code",
  "execution_count": 60,
  "id": "3e1ce831",
  "metadata": {},
  "outputs": [],
  "source": [
    "from sklearn.metrics import confusion_matrix, classification_report"
  ]
},
{
  "cell_type": "code",
  "execution_count": 61,
  "id": "1cf3b579",
  "metadata": {},
  "outputs": [
    {
      "data": {
        "text/plain": [
          "array([[2381,  33],\n        [ 436, 150]], dtype=int64)"
        ]
      },
      "execution_count": 61,
      "metadata": {},
      "output_type": "execute_result"
    }
  ],
  "source": [
    "confusion_matrix(y_test, y_pred)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 62,
  "id": "04053ee4",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "          precision    recall  f1-score   support\n",
        "\n",
        "    0      0.85      0.99      0.91     2414\n",
        "    1      0.82      0.26      0.39      586\n"
      ]
    }
  ]
}

```



```

        "\n",
        "    accuracy                0.84    3000\n",
        "    macro avg    0.83    0.62    0.65    3000\n",
        "    weighted avg    0.84    0.84    0.81    3000\n",
        "\n"
    ]
}
],
"source": [
    "print(classification_report(y_test, y_pred))"
]
},
{
    "cell_type": "markdown",
    "id": "ca48c401",
    "metadata": {},
    "source": [
        "# Hyperparameter Tunning"
    ]
},
{
    "cell_type": "code",
    "execution_count": 63,
    "id": "1338af67",
    "metadata": {},
    "outputs": [],
    "source": [
        "from sklearn.model_selection import GridSearchCV"
    ]
},
{
    "cell_type": "code",
    "execution_count": 64,
    "id": "c8da9523",
    "metadata": {},
    "outputs": [],
    "source": [
        "param_grid = {'C' : [0.1, 1, 10], 'gamma' : [1, 0.1, 0.01], 'kernel' : ['rbf'], 'class_weight' :
[balanced]}"
    ]
},
{
    "cell_type": "code",
    "execution_count": 65,
    "id": "27784340",
    "metadata": {},
    "outputs": [],
    "source": [

```

```

"grid = GridSearchCV(SVC(), param_grid, refit = True, verbose = 2, cv = 2)"
]
},
{
  "cell_type": "code",
  "execution_count": 66,
  "id": "dceebb46",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "Fitting 2 folds for each of 9 candidates, totalling 18 fits\n",
        "[CV] END ..C=0.1, class_weight=balanced, gamma=1, kernel=rbf; total time= 1.3s\n",
        "[CV] END ..C=0.1, class_weight=balanced, gamma=1, kernel=rbf; total time= 1.3s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.9s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.9s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.01, kernel=rbf; total time=
1.0s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.01, kernel=rbf; total time=
1.0s\n",
        "[CV] END ....C=1, class_weight=balanced, gamma=1, kernel=rbf; total time= 1.0s\n",
        "[CV] END ....C=1, class_weight=balanced, gamma=1, kernel=rbf; total time= 1.1s\n",
        "[CV] END ..C=1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.8s\n",
        "[CV] END ..C=1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.8s\n",
        "[CV] END .C=1, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.9s\n",
        "[CV] END .C=1, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.9s\n",
        "[CV] END ...C=10, class_weight=balanced, gamma=1, kernel=rbf; total time= 1.0s\n",
        "[CV] END ...C=10, class_weight=balanced, gamma=1, kernel=rbf; total time= 1.0s\n",
        "[CV] END .C=10, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.8s\n",
        "[CV] END .C=10, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.8s\n",
        "[CV] END C=10, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.8s\n",
        "[CV] END C=10, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.8s\n"
      ]
    },
    {
      "data": {
        "text/plain": [
          "GridSearchCV(cv=2, estimator=SVC(),\n",
          "              param_grid={'C': [0.1, 1, 10], 'class_weight': ['balanced']},\n",
          "              'gamma': [1, 0.1, 0.01], 'kernel': ['rbf']},\n",
          "              verbose=2)"
        ]
      },
      "execution_count": 66,
      "metadata": {},
      "output_type": "execute_result"
    }
  ]
}

```

```

    }
  ],
  "source": [
    "grid.fit(x_train, y_train)"
  ],
},
{
  "cell_type": "code",
  "execution_count": 67,
  "id": "fbdbc65b",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "SVC(C=10, class_weight='balanced', gamma=1)\n"
      ]
    }
  ],
},
  "source": [
    "print(grid.best_estimator_)"
  ],
},
{
  "cell_type": "code",
  "execution_count": 68,
  "id": "194cee17",
  "metadata": {},
  "outputs": [],
  "source": [
    "grid_predictions = grid.predict(x_test)"
  ],
},
{
  "cell_type": "code",
  "execution_count": 69,
  "id": "f2759f00",
  "metadata": {},
  "outputs": [
    {
      "data": {
        "text/plain": [
          "array([[2159, 255],\n       [ 343, 243]], dtype=int64)"
        ]
      }
    }
  ],
  "execution_count": 69,

```

```

    "metadata": {},
    "output_type": "execute_result"
  },
  "source": [
    "confusion_matrix(y_test, grid_predictions)"
  ],
  {
    "cell_type": "code",
    "execution_count": 70,
    "id": "ab7940f0",
    "metadata": {},
    "outputs": [
      {
        "name": "stdout",
        "output_type": "stream",
        "text": [
          "          precision    recall  f1-score   support\n",
          "\n",
          "   0      0.86      0.89      0.88     2414\n",
          "   1      0.49      0.41      0.45      586\n",
          "\n",
          " accuracy                0.80     3000\n",
          " macro avg      0.68      0.65      0.66     3000\n",
          "weighted avg      0.79      0.80      0.79     3000\n",
          "\n"
        ]
      }
    ],
    "source": [
      "print(classification_report(y_test, grid_predictions))"
    ]
  },
  {
    "cell_type": "markdown",
    "id": "af158ff3",
    "metadata": {},
    "source": [
      "# Model with Random Under Sampling"
    ]
  },
  {
    "cell_type": "code",
    "execution_count": 71,
    "id": "ddb101a1",
    "metadata": {},
    "outputs": [],

```

```

"source": [
  "svc_rus = SVC()"
]
},
{
  "cell_type": "code",
  "execution_count": 72,
  "id": "510e180f",
  "metadata": {},
  "outputs": [
    {
      "data": {
        "text/plain": [
          "SVC()"
        ]
      },
      "execution_count": 72,
      "metadata": {},
      "output_type": "execute_result"
    }
  ],
  "source": [
    "svc_rus.fit(x_train_rus, y_train_rus)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 73,
  "id": "769da95e",
  "metadata": {},
  "outputs": [],
  "source": [
    "y_pred_rus = svc_rus.predict(x_test_rus)"
  ]
},
{
  "cell_type": "markdown",
  "id": "58c7bb0c",
  "metadata": {},
  "source": [
    "# Model Accuracy"
  ]
},
{
  "cell_type": "code",
  "execution_count": 74,
  "id": "d4dbedfc",
  "metadata": {},

```

```

"outputs": [
  {
    "data": {
      "text/plain": [
        "array([[470, 157],\n",
        "       [174, 422]], dtype=int64)"
      ]
    },
    "execution_count": 74,
    "metadata": {},
    "output_type": "execute_result"
  }
],
"source": [
  "confusion_matrix(y_test_rus, y_pred_rus)"
]
},
{
  "cell_type": "code",
  "execution_count": 75,
  "id": "95246dac",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "          precision    recall  f1-score   support\n",
        "\n",
        "     0       0.73      0.75      0.74      627\n",
        "     1       0.73      0.71      0.72      596\n",
        "\n",
        "   accuracy                0.73    1223\n",
        "   macro avg      0.73      0.73      0.73    1223\n",
        "weighted avg      0.73      0.73      0.73    1223\n",
        "\n"
      ]
    }
  ],
  "source": [
    "print(classification_report(y_test_rus, y_pred_rus))"
  ]
},
{
  "cell_type": "markdown",
  "id": "16ecfe2f",
  "metadata": {},
  "source": [

```

```

"# Hyperparameter Tunning"
]
},
{
  "cell_type": "code",
  "execution_count": 76,
  "id": "4e84f039",
  "metadata": {},
  "outputs": [],
  "source": [
    "param_grid = {'C' : [0.1, 1, 10], 'gamma' : [1, 0.1, 0.01], 'kernel' : ['rbf'], 'class_weight' :
['balanced']}"
  ]
},
{
  "cell_type": "code",
  "execution_count": 77,
  "id": "6c5f0e36",
  "metadata": {},
  "outputs": [],
  "source": [
    "grid_rus = GridSearchCV(SVC(), param_grid, refit = True, verbose = 2, cv = 2)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 78,
  "id": "3e8f5290",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "Fitting 2 folds for each of 9 candidates, totalling 18 fits\n",
        "[CV] END ..C=0.1, class_weight=balanced, gamma=1, kernel=rbf; total time= 0.2s\n",
        "[CV] END ..C=0.1, class_weight=balanced, gamma=1, kernel=rbf; total time= 0.1s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.1s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.1s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.01, kernel=rbf; total time=
0.1s\n",
        "[CV] END C=0.1, class_weight=balanced, gamma=0.01, kernel=rbf; total time=
0.1s\n",
        "[CV] END ....C=1, class_weight=balanced, gamma=1, kernel=rbf; total time= 0.1s\n",
        "[CV] END ....C=1, class_weight=balanced, gamma=1, kernel=rbf; total time= 0.1s\n",
        "[CV] END ..C=1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.0s\n",
        "[CV] END ..C=1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.1s\n",
        "[CV] END .C=1, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.1s\n",

```

```

"[CV] END .C=1, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.1s\n",
"[CV] END ...C=10, class_weight=balanced, gamma=1, kernel=rbf; total time= 0.1s\n",
"[CV] END ...C=10, class_weight=balanced, gamma=1, kernel=rbf; total time= 0.1s\n",
"[CV] END .C=10, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.0s\n",
"[CV] END .C=10, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 0.1s\n",
"[CV] END C=10, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.0s\n",
"[CV] END C=10, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 0.1s\n"
]
},
{
  "data": {
    "text/plain": [
      "GridSearchCV(cv=2, estimator=SVC(),\n",
      "      param_grid={'C': [0.1, 1, 10], 'class_weight': ['balanced']},\n",
      "      'gamma': [1, 0.1, 0.01], 'kernel': ['rbf']},\n",
      "      verbose=2)"
    ]
  },
  "execution_count": 78,
  "metadata": {},
  "output_type": "execute_result"
},
{
  "source": [
    "grid_rus.fit(x_train_rus, y_train_rus)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 79,
  "id": "4737140f",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "SVC(C=1, class_weight='balanced', gamma=0.1)\n"
      ]
    }
  ],
  "source": [
    "print(grid_rus.best_estimator_)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 80,

```



```

"id": "541389d2",
"metadata": {},
"outputs": [],
"source": [
    "grid_pred_rus = grid_rus.predict(x_test_rus)"
]
},
{
    "cell_type": "code",
    "execution_count": 81,
    "id": "7e55bd4c",
    "metadata": {},
    "outputs": [
        {
            "data": {
                "text/plain": [
                    "array([[476, 151],\n",
                    "       [172, 424]], dtype=int64)"
                ]
            },
            "execution_count": 81,
            "metadata": {},
            "output_type": "execute_result"
        }
    ],
    "source": [
        "confusion_matrix(y_test_rus, grid_pred_rus)"
    ]
},
{
    "cell_type": "code",
    "execution_count": 82,
    "id": "76a4b497",
    "metadata": {},
    "outputs": [
        {
            "name": "stdout",
            "output_type": "stream",
            "text": [
                "      precision    recall  f1-score   support\n",
                "\n",
                "   0       0.73      0.76      0.75      627\n",
                "   1       0.74      0.71      0.72      596\n",
                "\n",
                " accuracy                0.74      1223\n",
                " macro avg       0.74      0.74      0.74      1223\n",
                "weighted avg       0.74      0.74      0.74      1223\n",
                "\n"
            ]
        }
    ]
}

```

```

    ]
  }
],
"source": [
  "print(classification_report(y_test_rus, grid_pred_rus))"
]
},
{
  "cell_type": "markdown",
  "id": "d5483903",
  "metadata": {},
  "source": [
    "# Model with Random Over Sampling"
  ]
},
{
  "cell_type": "code",
  "execution_count": 83,
  "id": "fcd3a270",
  "metadata": {},
  "outputs": [],
  "source": [
    "svc_ros = SVC()"
  ]
},
{
  "cell_type": "code",
  "execution_count": 84,
  "id": "d3b6de44",
  "metadata": {},
  "outputs": [
    {
      "data": {
        "text/plain": [
          "SVC()"
        ]
      },
      "execution_count": 84,
      "metadata": {},
      "output_type": "execute_result"
    }
  ],
  "source": [
    "svc_ros.fit(x_train_ros, y_train_ros)"
  ]
},
{
  "cell_type": "code",

```

```

"execution_count": 85,
"id": "2ccdbd6a",
"metadata": {},
"outputs": [],
"source": [
    "y_pred_ros = svc_ros.predict(x_test_ros)"
]
},
{
    "cell_type": "markdown",
    "id": "b0e950c6",
    "metadata": {},
    "source": [
        "# Model Accuracy"
    ]
},
{
    "cell_type": "code",
    "execution_count": 86,
    "id": "067c7443",
    "metadata": {},
    "outputs": [
        {
            "data": {
                "text/plain": [
                    "array([[1823, 556],\n       [ 626, 1773]], dtype=int64)"
                ]
            },
            "execution_count": 86,
            "metadata": {},
            "output_type": "execute_result"
        }
    ],
    "source": [
        "confusion_matrix(y_test_ros, y_pred_ros)"
    ]
},
{
    "cell_type": "code",
    "execution_count": 87,
    "id": "ccb0d59e",
    "metadata": {},
    "outputs": [
        {
            "name": "stdout",
            "output_type": "stream",
            "text": [

```

```

        precision recall f1-score support\n",
    "\n",
    "    0    0.74    0.77    0.76    2379\n",
    "    1    0.76    0.74    0.75    2399\n",
    "\n",
    " accuracy                0.75    4778\n",
    " macro avg    0.75    0.75    0.75    4778\n",
    "weighted avg    0.75    0.75    0.75    4778\n",
    "\n"
    ]
    }
    ],
    "source": [
        "print(classification_report(y_test_ros, y_pred_ros))"
    ]
    },
    {
        "cell_type": "markdown",
        "id": "f36ef9b3",
        "metadata": {},
        "source": [
            "# Hyperparameter Tunning"
        ]
    },
    {
        "cell_type": "code",
        "execution_count": 88,
        "id": "f4bb15fd",
        "metadata": {},
        "outputs": [],
        "source": [
            "param_grid = {'C' : [0.1, 1, 10], 'gamma' : [1, 0.1, 0.01], 'kernel' : ['rbf'], 'class_weight' :\n['balanced']}"
        ]
    },
    {
        "cell_type": "code",
        "execution_count": 89,
        "id": "f8287e69",
        "metadata": {},
        "outputs": [],
        "source": [
            "grid_ros = GridSearchCV(SVC(), param_grid, refit = True, verbose = 2, cv = 2)"
        ]
    },
    {
        "cell_type": "code",
        "execution_count": 90,

```

```

"id": "9011e315",
"metadata": {},
"outputs": [
  {
    "name": "stdout",
    "output_type": "stream",
    "text": [
      "Fitting 2 folds for each of 9 candidates, totalling 18 fits\n",
      "[CV] END ..C=0.1, class_weight=balanced, gamma=1, kernel=rbf; total time= 3.4s\n",
      "[CV] END ..C=0.1, class_weight=balanced, gamma=1, kernel=rbf; total time= 3.3s\n",
      "[CV] END C=0.1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 2.4s\n",
      "[CV] END C=0.1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 2.3s\n",
      "[CV] END C=0.1, class_weight=balanced, gamma=0.01, kernel=rbf; total time=
2.7s\n",
      "[CV] END C=0.1, class_weight=balanced, gamma=0.01, kernel=rbf; total time=
2.6s\n",
      "[CV] END ....C=1, class_weight=balanced, gamma=1, kernel=rbf; total time= 2.7s\n",
      "[CV] END ....C=1, class_weight=balanced, gamma=1, kernel=rbf; total time= 2.8s\n",
      "[CV] END ..C=1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 2.0s\n",
      "[CV] END ..C=1, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 2.0s\n",
      "[CV] END .C=1, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 2.3s\n",
      "[CV] END .C=1, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 2.2s\n",
      "[CV] END ...C=10, class_weight=balanced, gamma=1, kernel=rbf; total time= 2.5s\n",
      "[CV] END ...C=10, class_weight=balanced, gamma=1, kernel=rbf; total time= 2.5s\n",
      "[CV] END .C=10, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 2.1s\n",
      "[CV] END .C=10, class_weight=balanced, gamma=0.1, kernel=rbf; total time= 2.1s\n",
      "[CV] END C=10, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 2.2s\n",
      "[CV] END C=10, class_weight=balanced, gamma=0.01, kernel=rbf; total time= 2.1s\n"
    ]
  },
  {
    "data": {
      "text/plain": [
        "GridSearchCV(cv=2, estimator=SVC(),\n",
        "              param_grid={'C': [0.1, 1, 10], 'class_weight': ['balanced']},\n",
        "              'gamma': [1, 0.1, 0.01], 'kernel': ['rbf']},\n",
        "              verbose=2)"
      ]
    },
    "execution_count": 90,
    "metadata": {},
    "output_type": "execute_result"
  }
],
"source": [
  "grid_ros.fit(x_train_ros, y_train_ros)"
]
},

```

```

{
  "cell_type": "code",
  "execution_count": 91,
  "id": "d55881b1",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "SVC(C=10, class_weight='balanced', gamma=1)\n"
      ]
    }
  ],
  "source": [
    "print(grid_ros.best_estimator_)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 92,
  "id": "87bfddc6",
  "metadata": {},
  "outputs": [],
  "source": [
    "grid_pred_ros = grid_ros.predict(x_test_ros)"
  ]
},
{
  "cell_type": "code",
  "execution_count": 93,
  "id": "3b60b9bd",
  "metadata": {},
  "outputs": [
    {
      "data": {
        "text/plain": [
          "array([[2047, 332],\n        [ 68, 2331]], dtype=int64)"
        ]
      },
      "execution_count": 93,
      "metadata": {},
      "output_type": "execute_result"
    }
  ],
  "source": [
    "confusion_matrix(y_test_ros, grid_pred_ros)"
  ]
}

```

```

]
},
{
  "cell_type": "code",
  "execution_count": 94,
  "id": "406a5ca3",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "          precision    recall  f1-score   support\n",
        "\n",
        "         0      0.97      0.86      0.91     2379\n",
        "         1      0.88      0.97      0.92     2399\n",
        "\n",
        "    accuracy                0.92     4778\n",
        "   macro avg      0.92      0.92      0.92     4778\n",
        "weighted avg      0.92      0.92      0.92     4778\n",
        "\n"
      ]
    }
  ],
  "source": [
    "print(classification_report(y_test_ros, grid_pred_ros))"
  ]
},
{
  "cell_type": "markdown",
  "id": "44dedf84",
  "metadata": {},
  "source": [
    "# Lets Compare"
  ]
},
{
  "cell_type": "code",
  "execution_count": 95,
  "id": "74aaec6c",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "          precision    recall  f1-score   support\n",
        "\n",

```

```

"      0      0.85      0.99      0.91      2414\n",
"      1      0.82      0.26      0.39      586\n",
"\n",
" accuracy                0.84      3000\n",
" macro avg      0.83      0.62      0.65      3000\n",
"weighted avg      0.84      0.84      0.81      3000\n",
"\n"
]
}
],
"source": [
"print(classification_report(y_test, y_pred))"
]
},
{
"cell_type": "code",
"execution_count": 96,
"id": "f9dc6052",
"metadata": {},
"outputs": [
{
"name": "stdout",
"output_type": "stream",
"text": [
"      precision  recall  f1-score  support\n",
"\n",
"      0      0.86      0.89      0.88      2414\n",
"      1      0.49      0.41      0.45      586\n",
"\n",
" accuracy                0.80      3000\n",
" macro avg      0.68      0.65      0.66      3000\n",
"weighted avg      0.79      0.80      0.79      3000\n",
"\n"
]
}
],
"source": [
"print(classification_report(y_test, grid_predictions))"
]
},
{
"cell_type": "code",
"execution_count": 97,
"id": "c4c90355",
"metadata": {},
"outputs": [
{
"name": "stdout",

```



```

"output_type": "stream",
"text": [
  "      precision  recall f1-score  support\n",
  "\n",
  "      0      0.73    0.75    0.74    627\n",
  "      1      0.73    0.71    0.72    596\n",
  "\n",
  "  accuracy                0.73    1223\n",
  "  macro avg    0.73    0.73    0.73    1223\n",
  "weighted avg    0.73    0.73    0.73    1223\n",
  "\n"
]
},
],
"source": [
  "print(classification_report(y_test_rus, y_pred_rus))"
]
},
{
  "cell_type": "code",
  "execution_count": 98,
  "id": "b63889ae",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "      precision  recall f1-score  support\n",
        "\n",
        "      0      0.73    0.76    0.75    627\n",
        "      1      0.74    0.71    0.72    596\n",
        "\n",
        "  accuracy                0.74    1223\n",
        "  macro avg    0.74    0.74    0.74    1223\n",
        "weighted avg    0.74    0.74    0.74    1223\n",
        "\n"
      ]
    }
  ],
  "source": [
    "print(classification_report(y_test_rus, grid_pred_rus))"
  ]
},
{
  "cell_type": "code",
  "execution_count": 99,
  "id": "dce8dffe",

```

```

"metadata": {},
"outputs": [
  {
    "name": "stdout",
    "output_type": "stream",
    "text": [
      "      precision    recall  f1-score   support\n",
      "\n",
      "     0      0.74      0.77      0.76     2379\n",
      "     1      0.76      0.74      0.75     2399\n",
      "\n",
      "  accuracy                0.75     4778\n",
      " macro avg      0.75      0.75      0.75     4778\n",
      "weighted avg      0.75      0.75      0.75     4778\n",
      "\n"
    ]
  }
],
"source": [
  "print(classification_report(y_test_ros, y_pred_ros))"
]
},
{
  "cell_type": "code",
  "execution_count": 100,
  "id": "f9450c04",
  "metadata": {},
  "outputs": [
    {
      "name": "stdout",
      "output_type": "stream",
      "text": [
        "      precision    recall  f1-score   support\n",
        "\n",
        "     0      0.97      0.86      0.91     2379\n",
        "     1      0.88      0.97      0.92     2399\n",
        "\n",
        "  accuracy                0.92     4778\n",
        " macro avg      0.92      0.92      0.92     4778\n",
        "weighted avg      0.92      0.92      0.92     4778\n",
        "\n"
      ]
    }
  ],
  "source": [
    "print(classification_report(y_test_ros, grid_pred_ros))"
  ]
},

```

```
{
  "cell_type": "code",
  "execution_count": null,
  "id": "99502e33",
  "metadata": {},
  "outputs": [],
  "source": []
}
],
"metadata": {
  "kernelspec": {
    "display_name": "Python 3 (ipykernel)",
    "language": "python",
    "name": "python3"
  },
  "language_info": {
    "codemirror_mode": {
      "name": "ipython",
      "version": 3
    },
    "file_extension": ".py",
    "mimetype": "text/x-python",
    "name": "python",
    "nbconvert_exporter": "python",
    "pygments_lexer": "ipython3",
    "version": "3.8.12"
  }
},
"nbformat": 4,
"nbformat_minor": 5
}
```