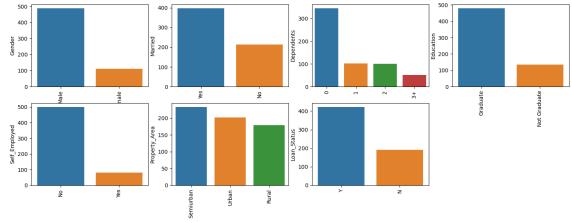
```
In [4]:
          M data.head
    Out[4]: <bound method NDFrame.head of
                                                     Loan ID
                                                               Gender Married Dependents
             Education Self_Employed \
             0
                   LP001002
                                Male
                                           No
                                                         0
                                                                Graduate
                                                                                       No
             1
                                Male
                   LP001003
                                          Yes
                                                         1
                                                                Graduate
                                                                                       No
             2
                   LP001005
                                Male
                                                         0
                                                                Graduate
                                          Yes
                                                                                      Yes
             3
                   LP001006
                                Male
                                          Yes
                                                         0
                                                            Not Graduate
                                                                                       No
             4
                                Male
                                                         0
                                                                Graduate
                   LP001008
                                           No
                                                                                       No
                                 . . .
                                          . . .
             609
                  LP002978
                              Female
                                           No
                                                         0
                                                                Graduate
                                                                                       No
             610
                  LP002979
                                Male
                                          Yes
                                                        3+
                                                                Graduate
                                                                                       No
             611
                   LP002983
                                Male
                                                         1
                                                                Graduate
                                          Yes
                                                                                       No
                                                         2
             612
                  LP002984
                                Male
                                          Yes
                                                                Graduate
                                                                                       No
             613
                  LP002990
                                           No
                                                         0
                                                                Graduate
                              Female
                                                                                      Yes
                   ApplicantIncome
                                      CoapplicantIncome
                                                           LoanAmount
                                                                        Loan_Amount_Term
             0
                               5849
                                                     0.0
                                                                   NaN
                                                                                     360.0
             1
                               4583
                                                  1508.0
                                                                 128.0
                                                                                     360.0
             2
                               3000
                                                     0.0
                                                                 66.0
                                                                                     360.0
             3
                               2583
                                                  2358.0
                                                                120.0
                                                                                    360.0
             4
                               6000
                                                     0.0
                                                                 141.0
                                                                                     360.0
                                . . .
                                                      . . .
                                                                   . . .
                                                                                       . . .
             . .
             609
                               2900
                                                     0.0
                                                                 71.0
                                                                                     360.0
             610
                               4106
                                                     0.0
                                                                 40.0
                                                                                    180.0
             611
                                                   240.0
                                                                 253.0
                                                                                    360.0
                               8072
             612
                                                     0.0
                                                                 187.0
                                                                                    360.0
                               7583
             613
                               4583
                                                     0.0
                                                                133.0
                                                                                    360.0
                   Credit_History Property_Area Loan_Status
             0
                                            Urban
                               1.0
                                                              Υ
                               1.0
                                            Rural
                                                              N
             1
             2
                               1.0
                                            Urban
                                                              Υ
             3
                               1.0
                                            Urban
                                                              Υ
             4
                                                              Υ
                               1.0
                                            Urban
                               . . .
                                               . . .
                               1.0
                                                              Υ
             609
                                            Rural
             610
                               1.0
                                            Rural
                                                              Υ
             611
                               1.0
                                            Urban
                                                              Υ
             612
                                            Urban
                                                              Υ
                               1.0
             613
                               0.0
                                        Semiurban
             [614 rows x 13 columns]>
In [5]:
             obj = (data.dtypes == 'object')
             print("Categorical variables:",len(list(obj[obj].index)))
             Categorical variables: 8
             data.drop(['Loan_ID'],axis=1,inplace=True)
In [6]:
```

```
In [7]: N

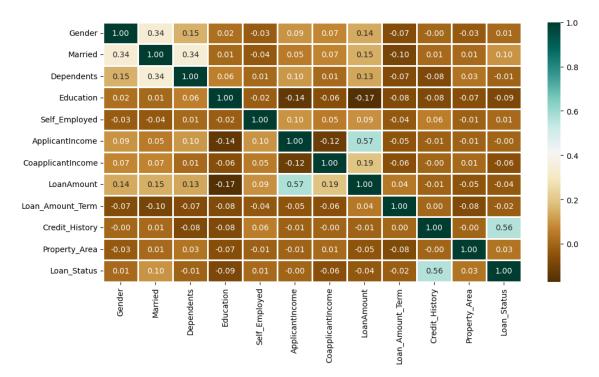
obj = (data.dtypes == 'object')
object_cols = list(obj[obj].index)
plt.figure(figsize=(18,36))
index = 1

for col in object_cols:
    y = data[col].value_counts()
    plt.subplot(11,4,index)
    plt.xticks(rotation=90)
    sns.barplot(x=list(y.index), y=y)
    index +=1
```



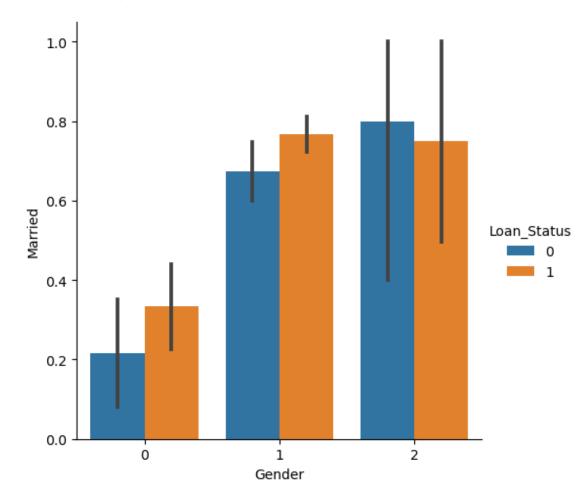
Categorical variables: 0

Out[10]: <Axes: >



```
In [11]: N
sns.catplot(x="Gender", y="Married",
hue="Loan_Status",
kind="bar",
data=data)
```

Out[11]: <seaborn.axisgrid.FacetGrid at 0x169e3451050>



```
for col in data.columns:
In [12]:
               data[col] = data[col].fillna(data[col].mean())
             data.isna().sum()
   Out[12]: Gender
                                   0
             Married
                                   0
             Dependents
                                   0
             Education
             Self_Employed
                                   0
             ApplicantIncome
             CoapplicantIncome
                                   0
             LoanAmount
                                   0
             Loan_Amount_Term
                                   0
             Credit_History
                                   0
                                   0
             Property_Area
                                   0
             Loan_Status
```

dtype: int64

```
In [13]:
             from sklearn.model selection import train test split
             X = data.drop(['Loan_Status'],axis=1)
             Y = data['Loan_Status']
             X.shape, Y.shape
             X_train, X_test, Y_train, Y_test = train_test_split(X, Y,
                                                                 test_size=0.4,
                                                                  random_state=1)
             X_train.shape, X_test.shape, Y_train.shape, Y_test.shape
   Out[13]: ((368, 11), (246, 11), (368,), (246,))
In [14]:
             from sklearn.neighbors import KNeighborsClassifier
             from sklearn.ensemble import RandomForestClassifier
             from sklearn.svm import SVC
             from sklearn.linear_model import LogisticRegression
             from sklearn import metrics
             knn = KNeighborsClassifier(n_neighbors=3)
             rfc = RandomForestClassifier(n_estimators = 7,
                                          criterion = 'entropy',
                                          random_state =7)
             svc = SVC()
             lc = LogisticRegression()
             # making predictions on the training set
             for clf in (rfc, knn, svc,lc):
                 clf.fit(X_train, Y_train)
                 Y_pred = clf.predict(X_train)
                 print("Accuracy score of ",
                       clf.__class__.__name___,
                       "=",100*metrics.accuracy_score(Y_train,
                                                      Y pred))
             Accuracy score of RandomForestClassifier = 97.01086956521739
             Accuracy score of KNeighborsClassifier = 77.17391304347827
             Accuracy score of SVC = 70.38043478260869
             Accuracy score of LogisticRegression = 82.88043478260869
             C:\Users\lenovo\anaconda3\Lib\site-packages\sklearn\linear_model\_logisti
             c.py:458: ConvergenceWarning: lbfgs failed to converge (status=1):
             STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
             Increase the number of iterations (max_iter) or scale the data as shown i
                 https://scikit-learn.org/stable/modules/preprocessing.html (https://s
             cikit-learn.org/stable/modules/preprocessing.html)
             Please also refer to the documentation for alternative solver options:
                 https://scikit-learn.org/stable/modules/linear_model.html#logistic-re
             gression (https://scikit-learn.org/stable/modules/linear model.html#logis
             tic-regression)
               n_iter_i = _check_optimize_result(
```

```
    for clf in (rfc, knn, svc,lc):

In [15]:
                 clf.fit(X_train, Y_train)
                 Y_pred = clf.predict(X_test)
                 print("Accuracy score of ",
                       clf.__class__.__name__,"=",
                       100*metrics.accuracy_score(Y_test,
                                                  Y_pred))
             Accuracy score of RandomForestClassifier = 76.42276422764228
             Accuracy score of KNeighborsClassifier = 62.19512195121951
             Accuracy score of SVC = 67.07317073170732
             Accuracy score of LogisticRegression = 78.86178861788618
             C:\Users\lenovo\anaconda3\Lib\site-packages\sklearn\linear_model\_logisti
             c.py:458: ConvergenceWarning: lbfgs failed to converge (status=1):
             STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
             Increase the number of iterations (max_iter) or scale the data as shown i
                 https://scikit-learn.org/stable/modules/preprocessing.html (https://s
             cikit-learn.org/stable/modules/preprocessing.html)
             Please also refer to the documentation for alternative solver options:
                 https://scikit-learn.org/stable/modules/linear_model.html#logistic-re
             gression (https://scikit-learn.org/stable/modules/linear model.html#logis
             tic-regression)
               n_iter_i = _check_optimize_result(
 In [ ]:
In [ ]:
```