

# Régression Multinomiale - Classification Multiclasse

## Régression Multinomiale (Softmax Regression)

### Théorie

La **régression multinomiale**, aussi appelée **régression logistique multinomiale**, est une extension de la régression logistique qui permet de gérer plusieurs classes. Elle utilise une fonction **Softmax** en sortie pour assigner une probabilité à chaque classe.

### Hyperparamètres

Nous allons optimiser un seul hyperparamètre pour réduire le temps d'entraînement : - **Paramètre de régularisation (C)** : contrôle la pénalisation de la complexité du modèle (valeurs entre 0.1 et 1).

### Exemple en Python

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler

# Chargement des ensembles de données
train_data = pd.read_csv('covertime_train.csv')
val_data = pd.read_csv('covertime_val.csv')
```

```

test_data = pd.read_csv('coverttype_test.csv')

# Préparation des données
X_train = train_data.drop('Cover_Type', axis=1)
y_train = train_data['Cover_Type']

X_val = val_data.drop('Cover_Type', axis=1)
y_val = val_data['Cover_Type']

X_test = test_data.drop('Cover_Type', axis=1)
y_test = test_data['Cover_Type']

# Standardisation des données
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_val = scaler.transform(X_val)
X_test = scaler.transform(X_test)

# Recherche du meilleur hyperparamètre (C seulement)
C_values = [0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1] # Entre 0.1 et 1
val_accuracies = []

for C in C_values:
    model = LogisticRegression(multi_class='multinomial', solver='saga', C=C, penalty='l2', n_iter=1000)
    model.fit(X_train, y_train)
    y_val_pred = model.predict(X_val)
    acc = accuracy_score(y_val, y_val_pred)
    val_accuracies.append((C, acc))

# Sélection du meilleur hyperparamètre
best_C, best_acc = max(val_accuracies, key=lambda x: x[1])
print(f"Meilleur hyperparamètre : C={best_C}")

# Affichage du graphique
plt.figure(figsize=(8, 6))
plt.plot(C_values, [acc for C, acc in val_accuracies], marker='o', linestyle='dashed')
plt.xlabel("Paramètre de régularisation (C)")
plt.ylabel("Précision sur validation")
plt.title("Impact de la régularisation sur la performance de la régression multinomiale")
plt.show()

# Modèle final avec le meilleur hyperparamètre

```

```

final_model = LogisticRegression(multi_class='multinomial', solver='saga', C=best_C, penalty=
final_model.fit(X_train, y_train)
y_test_pred = final_model.predict(X_test)

# Affichage de la matrice de confusion
conf_matrix = confusion_matrix(y_test, y_test_pred)
print("\nMatrice de confusion :")
print(conf_matrix)

print("\nÉvaluation sur l'ensemble de test")
print(classification_report(y_test, y_test_pred))

```

```

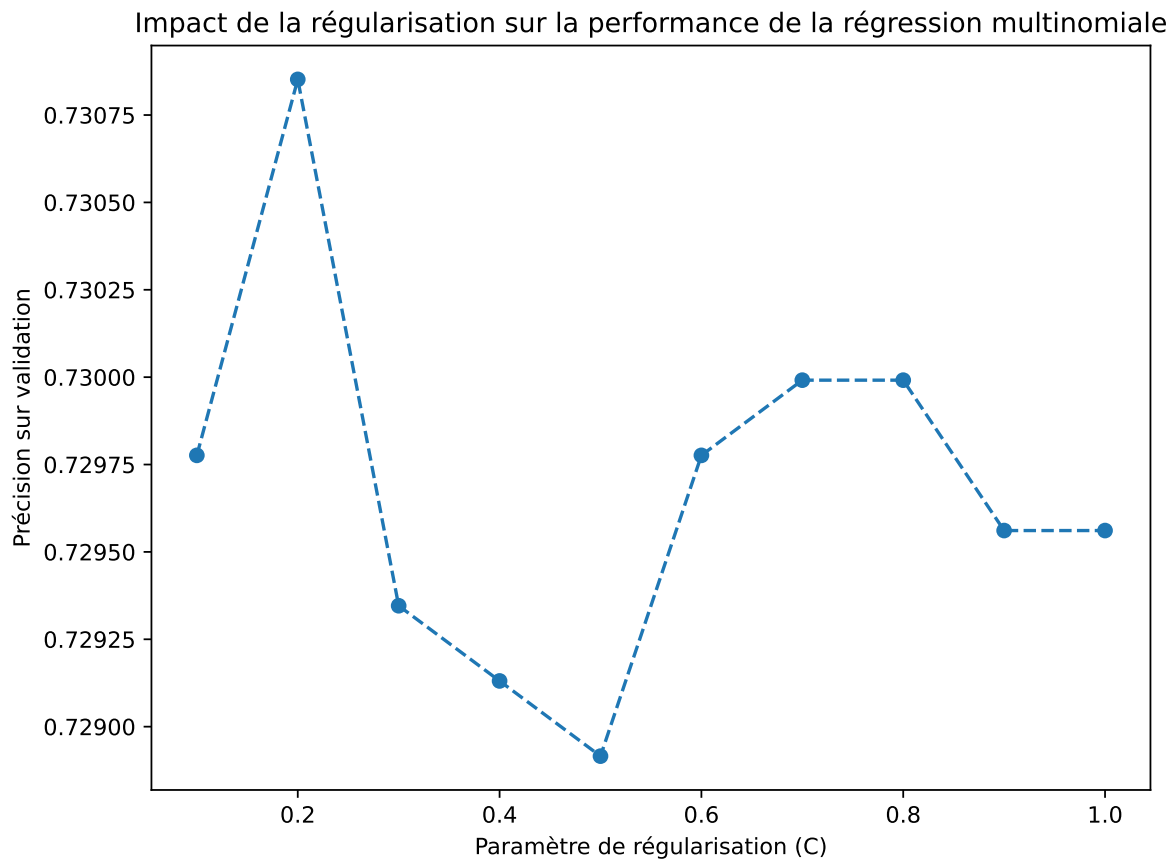
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: FutureWarning:
  warnings.warn(

```

```
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: Convergence
warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: Futur
warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: Convergence
warnings.warn(
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: Futur
warnings.warn(
```

Meilleur hyperparamètre : C=0.2

```
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: Convergence
warnings.warn(
```



```
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_logistic.py:1247: Futur
warnings.warn(
```

Matrice de confusion :

```
[[1206 471  0  0  0  1 30]
 [ 415 1790 34  0  1 20  1]
 [  0  27 226  7  0 21  0]
 [  0  0  7 10  0  4  0]
 [  1  68  4  0  0  1  0]
 [  0  35 84  1  0 24  0]
 [ 71  0  0  0  0  0 88]]
```

Évaluation sur l'ensemble de test

	precision	recall	f1-score	support
1	0.71	0.71	0.71	1708
2	0.75	0.79	0.77	2261
3	0.64	0.80	0.71	281
4	0.56	0.48	0.51	21
5	0.00	0.00	0.00	74
6	0.34	0.17	0.22	144
7	0.74	0.55	0.63	159
accuracy			0.72	4648
macro avg	0.53	0.50	0.51	4648
weighted avg	0.70	0.72	0.71	4648

```
/home/ensai/.local/lib/python3.10/site-packages/sklearn/linear_model/_sag.py:349: ConvergenceWarning:
  warnings.warn(
```