

Assignment 2 - Interpolation

March 18, 2025

1 Assignment 2: Interpolation de Lagrange

On considère la fonction

$$f(x) = 1 - h(g(x)) \quad \text{avec} \quad h(y) = \frac{1}{1 - 2y + 4y^2} \quad \text{et} \quad g(x) = \frac{x^2 - 4}{4}$$

sur l'intervalle $I = [-3, 3]$.

1.1 Partie 1

1.1.1 Partie 1a

Calculer le polynôme d'interpolation $\Pi_n f(x)$ de degré $n = 20$

- avec des noeuds équidistribués ;
- avec les noeuds de Chebyshev ;

et comparer graphiquement les résultats obtenus avec la fonction donnée. Est-ce que l'interpolation est toujours précise? Pourquoi? Y aurait-il des différences si nous utilisions l'interpolation polynômiale à l'aide de la matrice de Vandermonde au lieu de l'interpolation de Lagrange? (*Réponse 1*)

```
[ ]: import numpy as np
import matplotlib.pyplot as plt
```

```
[ ]: # HELPER FUNCTIONS FOR LAGRANGE INTERPOLATION -- THESE ARE GIVEN TO YOU,
↪NOTHING TO COMPLETE !
```

```
def LagrangeBasis(t,k,z):
    """Evaluate the 'k'-th Lagrange basis function at 'z',
    given the interpolatory points 't'.
    INPUT:
        t : array of interpolatory points
        k : number of the basis function
        z : array of points where to evaluate the basis
    OUTPUT:
        result: evaluation of the Lagrange basis function at 'z'
    """

    n = t.shape[0] - 1
```

```

# init result to one, of same type and size as z
result = np.zeros_like(z) + 1

# first few checks on k:
if (type(k) is not int) or (t.shape[0] < 1) or (k > n) or (k < 0):
    raise ValueError('Lagrange basis needs a positive integer k, smaller_
↳ than the size of t')

# loop on n to compute the product
for j in range(n+1) :
    if (j == k) :
        continue
    if (t[k] == t[j]) :
        raise ValueError('All the interpolation points need to be distinct')

    result *= (z - t[j]) / (t[k] - t[j])

return result

def LagrangeInterpolation(t,y,z):
    """Evaluate the Lagrange interpolant function at 'z',
    given the interpolatory points 't' and the data 'y'.
    INPUT:
        t : array of interpolatory points
        y : array of data to be interpolated
        z : array of points where to evaluate the interpolant
    OUTPUT:
        result: evaluation of the interpolant function at 'z'
    """

    # {phi(t,k,.), k=0,...,n} is a basis of the polynomials of degree n
    # y represents the coordinates of the interpolating polynomial with respect_
↳ to this basis.
    # Therefore LagrangePi(t,y,.) = y[0] phi(t,0,.) + ... + y[n] phi(t,n,.)

    n = t.size - 1

    # init result to zero, of same type and size as z
    result = np.zeros_like(z)

    # loop on n to compute the product
    for k in range(t.shape[0]) :
        result += y[k] * LagrangeBasis(t,k,z)

    return result

```

```
[ ]: # definition of the function to be interpolated
g1 = lambda x : (x**2 - 4) / 4
h = lambda y : 1 / (1 - 2*y + 4*y**2)
f = lambda x : 1 - h(g1(x))
a, b = -3, 3
```

```
[ ]: # EQUIDISTRIBUTED NODES - PLOT
x = np.linspace(a,b,1000) # fine partition, for nice plotting
n=20

xp = ## COMPLETE HERE !! ## (equispaced nodes in [a,b])
fn = ## COMPLETE HERE !! ## (interpolant polynomial)

plt.figure(figsize=(10,6))
## COMPLETE HERE !! ## (plot the interpolant polynomial)
plt.plot(x, f(x), 'k')
plt.plot(xp, f(xp), 'ok')
plt.xlabel(r'$\mathbf{x}$', fontsize=20)
plt.ylabel(r'$\mathbf{f(x)}$', fontsize=20)
plt.legend([f"Lagrange Equi n={n}", '$ f(x)$'], fontsize=15)
plt.grid()
plt.title(f"Interpolation de Lagrange avec noeuds équirépartis - n={n}",
          fontsize=15, fontweight="bold")
plt.show()
```

```
[ ]: # CHEBYSHEV NODES - PLOT
x = np.linspace(a,b,1000) # fine partition, for nice plotting
n=20

# COMPLETE HERE : define xc as the Chebyshev nodes in [a,b]. (few lines of code)
# xc =

fn = ## COMPLETE HERE !! ## (interpolant polynomial)

plt.figure(figsize=(10,6))
## COMPLETE HERE !! ## (plot the interpolant polynomial)
plt.plot(x, f(x), 'k')
plt.plot(xc, f(xc), 'ok')
plt.xlabel(r'$\mathbf{x}$', fontsize=20)
plt.ylabel(r'$\mathbf{f(x)}$', fontsize=20)
plt.legend([f"Lagrange Chebyshev n={n}", '$ f(x)$'], fontsize=15)
plt.grid()
plt.title(f"Interpolation de Lagrange avec noeuds de Chebyshev - n={n}",
          fontsize=15, fontweight="bold")
plt.show()
```

1.1.2 Commentaire

Réponse 1 Ecrivez ici votre reponse

1.1.3 Partie 1b

Évaluer les erreurs d'interpolation

$$E_n^a(f) = \max_{x \in I} |f(x) - \Pi_n f(x)| \quad (\text{erreur absolue}) \qquad E_n^r(f) = \max_{x \in I} \frac{|f(x) - \Pi_n f(x)|}{|f(x)|} \quad (\text{erreur relative})$$

Visualiser le graphe logarithmique de l'erreur absolue E_n^a en fonction de n pour les noeuds équirépartis et pour les noeuds de Chebyshev, avec $n = 5, 10, 15, \dots, 50$. Est-ce que les résultats sont en accord avec la théorie vue au cours ? (*Réponse 2*)

```
[ ]: # EQUIDISTRIBUTED NODES - ERRORS
Nrange = np.arange(5,51,5)
errorLag = []
errorLagRel = []
for n in Nrange:
    xp = ## COMPLETE HERE !! ## (equispaced nodes in [a,b])
    fn = ## COMPLETE HERE !! ## (interpolant polynomial)
    errorLag.append() ## COMPLETE HERE !! # (absolute error)
    errorLagRel.append() ## COMPLETE HERE !! # (relative error)

plt.figure(figsize=(10,6))
plt.plot(Nrange, errorLag, ':o');
plt.yscale('log')
plt.xlabel('Degree', fontsize=20, fontweight='bold')
plt.ylabel('Error', fontsize=20, fontweight='bold')
plt.grid()
plt.show()
```

```
[ ]: # CHEBYSHEV NODES - ERRORS
Nrange = np.arange(5,51,5)
errorChe = []
errorCheRel = []
for n in Nrange:
    xc = ## COMPLETE HERE !! ## (Chebyshev nodes in [a,b], as before)
    fn = ## COMPLETE HERE !! ## (interpolant polynomial)
    errorChe.append() ## COMPLETE HERE !! # (absolute error)
    errorCheRel.append() ## COMPLETE HERE !! # (relative error)

plt.figure(figsize=(10,6))
plt.plot(Nrange, errorChe, ':o')
plt.yscale('log')
plt.xlabel('Degree', fontsize=20, fontweight='bold')
plt.ylabel('Error', fontsize=20, fontweight='bold')
```

```
plt.grid()
plt.show()
```

1.1.4 Commentaire

Réponse 2 Ecrivez ici votre réponse

1.2 Partie 2

Écrire une fonction `PiecewiseInterpolation`, qui implemente l'interpolation par intervalles. La fonction a la structure suivante

```
def PiecewiseInterpolation(N, LocalOrder, a, b, f, z):
    """
    This function implements piecewise interpolation considering (1) a user-defined number of
    subintervals, and (2) a user-defined polynomial degree in each subinterval.

    INPUTS:
    N: the number of subintervals
    LocalOrder : order of the polynomial in each subinterval
    a,b : global extrema of the interval
    f : the values of the function at the interpolation nodes
    z : where to evaluate the function

    OUTPUTS:
    result: the evaluation of the piecewise interpolant of f at z
    """
```

```
[ ]: from collections.abc import Iterable

def PiecewiseInterpolation(N, LocalOrder, a, b, f, z):
    """
    This function implements piecewise interpolation considering (1) a
    user-defined number of subintervals, and
    (2) a user-defined polynomial degree in each subinterval.

    INPUTS:
    N: the number of subintervals
    LocalOrder : order of the polynomial in each subinterval
    a,b : global extrema of the interval
    f : the values of the function at the interpolation nodes
    z : where to evaluate the function

    OUTPUTS:
    result: the evaluation of the piecewise interpolant of f at z
    """

    def localInterpolation(zk):
        """Interpolates the function f at the single point zk
```

```

"""

# find out in which interval lies the point zk
i = 0
for k in range(x.shape[0] - 1):
    if x[k] <= zk <= x[k+1]:
        break
    else:
        i += 1

# hereafter is an alternative code to detect the interval
# i = np.where([x[k] <= zk <= x[k+1] for k in range(x.
↪shape[0]-1)))[0][0]

# if zk is not in the interval, return constant function
if zk == x[i]:
    return f(x[i])
elif zk == x[i+1]:
    return f(x[i+1])

# otherwise, compute the local interpolation nodes in the i-th interval
x_loc = ## COMPLETE HERE !! ## (local equispaced nodes)

return ## COMPLETE HERE !! ## (evaluation of the interpolant at zk)

H = (b-a)/N
if np.isclose(H,0) :
    raise ValueError('PiecewiseInterpolation : a and b are too close!')

# Intervals of constant length
x = ## COMPLETE HERE !! ## (points that define the sub-intervals)

if not isinstance(z, Iterable):
    z = [z]

result = np.zeros_like(z)
for k, zk in enumerate(z) :
    result[k] = ## COMPLETE HERE !! ## (interpolate f at the current point)↪
↪

return result

```

1.3 Partie 3

Considérer des noeuds équidistribués.

Calculer les polynômes interpolatoires par intervalles d'ordre 1,2,3 (c-à-d $\Pi_1^H f(x)$, $\Pi_2^H f(x)$, $\Pi_3^H f(x)$) sur N sous-intervalles de longueur $H = \frac{b-a}{N}$, avec $N = 5, 15, 30, 60$ et comparer graphiquement les

résultats obtenus avec la fonction donnée.

Prendre $N = 5, 10, 15, \dots, 50$ (nombre de sous-intervalles). Calculer les polynômes interpolatoires par intervalles d'ordre 1,2,3 et évaluer les erreurs absolue et relative (cf. Section 1). Visualiser les graphes logarithmiques des erreurs absolues E_H^a en fonction de N . Est-ce que l'erreur diminue en utilisant l'interpolation par intervalles? Pourquoi? (*Réponse 3*)

```
[ ]: localOrders = [1,2,3]
```

```
[ ]: # plots of piecewise interpolants of (local) degree 1,2,3

fig, axs = plt.subplots(1, 3, figsize=(30,10), sharey=True)

Nrangeplot = [5,15,30,60]
for cnt_o,order in enumerate(localOrders):
    for n in Nrangeplot:
        fn = ## COMPLETE HERE !! ## (piecewise interpolant polynomial of f of
        ↪degree 'order')
        axs[cnt_o].plot(x, fn, ':', label=f"N={n}", linewidth=2)

for ax in axs:
    ax.plot(x, f(x), 'k', label="f(x)", linewidth=3)
    ax.legend(fontsize=20)
    ax.set_xlabel(r'$\mathbf{x}$', fontsize=20)
    ax.set_ylabel(r'$\mathbf{f(x)}$', fontsize=20)
    ax.grid(visible=True, which='major', color='#666666', linestyle='-')
    ax.minorticks_on()
    ax.grid(visible=True, which='minor', color='#999999', linestyle='-',alpha=0.
    ↪2)

plt.show()
```

```
[ ]: # absolute error trends of piecewise interpolants of (local) degree 1,2,3

Nrange = np.arange(5,51,5)
errorPwi = np.zeros((len(localOrders), len(Nrange)))
errorPwiRel = np.zeros((len(localOrders), len(Nrange)))
for cnt_o,order in enumerate(localOrders):
    for cnt_n,n in enumerate(Nrange):
        fn = ## COMPLETE HERE !! ## (piecewise interpolant polynomial of f of
        ↪degree 'order')
        errorPwi[cnt_o, cnt_n] = ## COMPLETE HERE !! ## (absolute error)
        errorPwiRel[cnt_o, cnt_n] = ## COMPLETE HERE !! ## (relative error)

plt.figure(figsize=(10,6))
for cnt_o,order in enumerate(localOrders):
    plt.semilogy(Nrange, errorPwi[cnt_o], ':o', label=f"order {order}")
plt.xlabel('Nombre de sous-intervalles', fontsize=20, fontweight='bold')
```

```
plt.ylabel('Error', fontsize=20, fontweight='bold')

plt.grid(visible=True, which='major', color='#666666', linestyle='-')
plt.minorticks_on()
plt.grid(visible=True, which='minor', color='#999999', linestyle='-', alpha=0.2)
plt.legend(fontsize=15)
plt.ylim([1e-4, None])
plt.show()
```

1.3.1 Commentaire

Réponse 3 Ecrivez ici votre reponse

1.4 Partie 4

Finalement, parmi les méthodes analysées, laquelle nous permet d'interpoler la fonction donnée avec une **erreur relative** au plus de 10%, en minimisant le nombre de points où il faut évaluer la fonction f ? **En regardant simplement la figure**, justifiez votre réponse. (*Réponse 4*) (*Réponse 4*)

```
[ ]: plt.figure(figsize=(10,6))

#plt.plot(Nrange, errorLagRel, label = 'Lagrange')
plt.plot(Nrange, errorCheRel, '-v', label='Chebyshev')
for cnt_o, order in enumerate(localOrders):
    plt.plot(Nrange, errorPwiRel[cnt_o], '-o', label=f'PWI - Order {order}')

tol = 0.1
plt.plot(Nrange, tol*np.ones(len(Nrange)), 'k-.', label='$10 \\\$ threshold',
         linewidth=2)

plt.yscale('log')
plt.xlabel('n or N', fontsize=20, fontweight='bold')
plt.ylabel('Error', fontsize=20, fontweight='bold')
plt.grid(which='major', linestyle='-', linewidth=1)
plt.minorticks_on()
plt.grid(which='minor', color='#999999', linestyle='--', alpha=0.2)
plt.legend(fontsize=12, bbox_to_anchor=(0.975, 0.975), borderaxespad=0)
plt.show()
```

1.4.1 Commentaire

Réponse 4 Ecrivez ici votre reponse

2 Quelques petites questions finales (pas évaluées)

- What types of collaboration strategies did your group use?

- Work in pairs on different sections.
 - Work individually on different sections.
 - Work together on the same section with one notebook opened.
 - Work together on the same section with multiple notebooks opened.
 - Other (please specify).
- How effective was your collaboration strategy today? Please rate from 1 (not at all) to 5 (very effective).
 - How supported did you feel by your TA during the session today? Please rate from 1 (not at all) to 5 (very effective).

Please report your answers here. **Thank you!**