

# Polynomial Translation

Given an  $n^{th}$  degree polynomial,  $p$ , it's translation by a horizontal offset,  $h \neq 0$ , can be expressed as a new  $n^{th}$  degree polynomial  $q$ ,

$$p(x + h) = q(x),$$

where,

$$p(x) = \sum_{i=0}^n a_i x^i,$$

$$q(x) = \sum_{i=0}^n b_i x^i,$$

$$b_i = \sum_{j=i}^n \binom{j}{i} (-h)^{j-i} a_j$$

## This comes directly from the binomial theorem

Expanding  $p(x + h)$  using binomial theorem and reversing summation indices,

$$p(x + h) = \sum_{i=0}^n a_i (x + h)^i = \sum_{i=0}^n \sum_{j=0}^i a_i x^j \binom{i}{j} h^{i-j} = \sum_{i=0}^n x^i \sum_{j=i}^n \binom{j}{i} h^{j-i} a_j$$

## Represented as a system...

Using an  $(n + 1) \times (n + 1)$  matrix,  $D_h$ , for  $q$ 's representation,

$$\vec{p} = \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix}, \quad \vec{q} = \begin{pmatrix} b_0 \\ b_1 \\ \vdots \\ b_n \end{pmatrix}, \quad D_h = \begin{pmatrix} d_{00} & d_{01} & \dots & d_{0n} \\ 0 & d_{11} & \dots & d_{1n} \\ \vdots & \vdots & d_{ij} & \vdots \\ 0 & 0 & \dots & d_{nn} \end{pmatrix}, \quad d_{ij} = \binom{j}{i} (-h)^{j-i}$$

defining  $\binom{j}{i}$  as,

$$\binom{j}{i} = \begin{cases} \frac{j!}{(j-i)!i!}, & \text{if } j \geq i \geq 0 \\ 0, & \text{otherwise} \end{cases}$$

Then the coefficients for the translated polynomial  $p(x + h)$  or  $r(x)$  can be found by solving the following system (as demonstrated bellow in code),

$$D_h \vec{p} = \vec{q} \tag{2}$$

In [140]:

```
import random

import numpy as np
from numpy.polynomial.polynomial import polyval
from numpy.linalg import matrix_power
from numpy.linalg import matrix_rank

from scipy.linalg import lu
from scipy.linalg import pascal

import matplotlib.pyplot as plt

import sympy
import sympy as sp
from sympy.abc import x

from IPython.display import display, Math, Latex

#
# Display Numpy Poly1d's
#

def plot_polys(polys, a=-10, b=10, steps=100):
    sp.init_printing()
    from sympy.abc import x
    X = np.linspace(a,b,steps)

    for p in polys:
        y = p(X)
        l = sp.Poly(p.c,x).as_expr()
        plt.plot(X,y, label=l)

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

```

#
# create Horizontal Shift Matrix
#

def poly_shift_matrix(order, h):
    # h -- horizontal shift amount

    # n+1 coefficients to solve for -- n+1 x n+1 shift matrix
    n = order + 1

    # pascal matrix
    P = pascal(n, kind='upper')

    # create H
    H = np.eye(n)
    _h = 1
    for k in range(n):
        for i in range(n-k):
            H[i,i+k] = _h
        _h *= h

    # create shift matrix - flip for numpy convention
    S = np.flip(H * P)

    # return
    return S

```

```

#
# Horizontal Shift
#

```

```

def poly_horizontal_translation(p, delta_x):

    #  $r(x) = p(x + \text{delta\_x})$ 
    S = poly_shift_matrix(p.order,delta_x)
    r = np.polyld( S @ p )

    # return  $r(x)$  -- numpy polynomial
    return r

```

```

#
# Vertical Shift
#

```

```

def poly_vertical_translation(poly, delta_y):
    #  $r(x) = p(x) + \text{delta\_y}$ 
    r = poly + delta_y
    return r

```

```

#
# Translate Numpy Polynomial
#

```

```

#
def translate_polynomial(p, delta_x=0, delta_y=0):
    r = p

    if delta_x != 0:
        #  $r(x) = p(x + \text{delta\_x})$ 
        r = poly_horizontal_translation(r, delta_x)

    if delta_y != 0:
        #  $r(x) = p(x) + \text{delta\_y}$ 
        r = poly_vertical_translation(r, delta_y)

    return r

#
# Random Testing
#

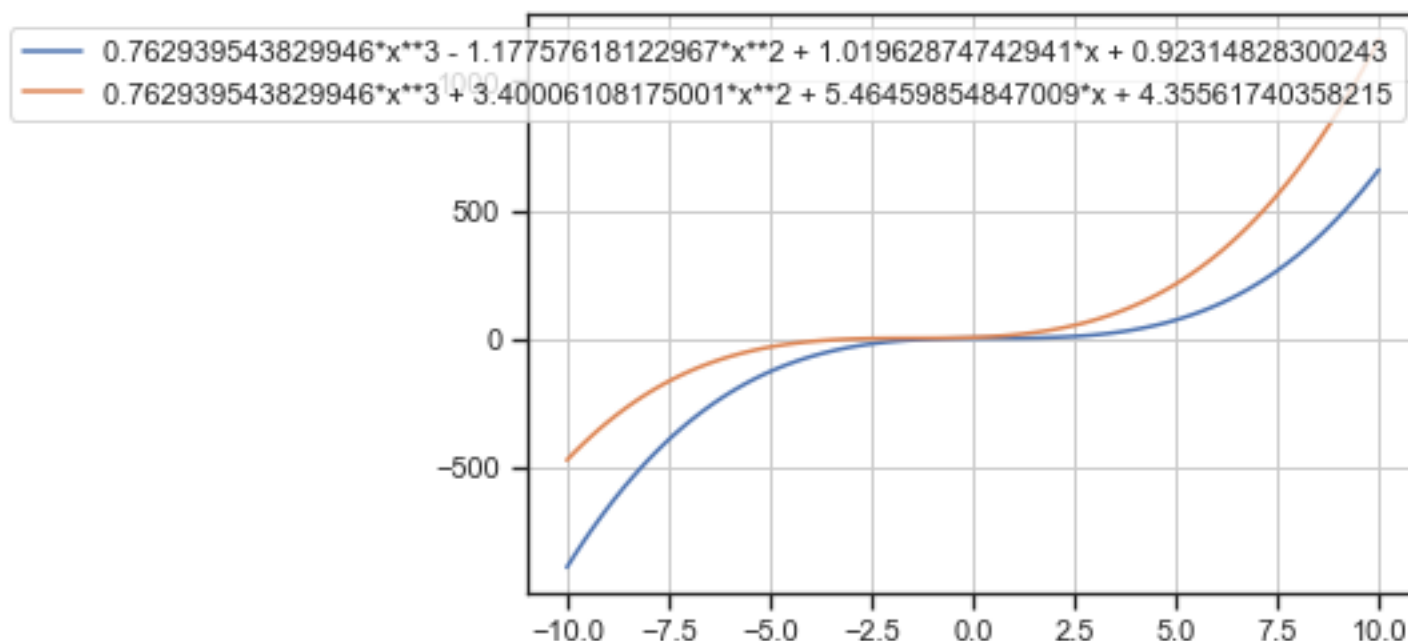
# SETTINGS
MIN_DEGREE = 2
MAX_DEGREE = 5
SHIFT_RANGE = 2 # MAX

# generate random base polynomial
n = random.randint(MIN_DEGREE, MAX_DEGREE)
p = np.poly1d( np.random.randn(n))

# translate base polynomial
x_shift, y_shift = random.randint(-SHIFT_RANGE, SHIFT_RANGE), random.randint(-SHIFT_RANGE, SHIFT_RANGE)
q = translate_polynomial(p, delta_x=x_shift, delta_y=y_shift)

# plot results
plot_polys([p, q], a=-10, b=10)

```



# Interpolation of recursively defined sequences

One interesting use case of the translation Matrix,  $D_h$ , defined above involves finding an interpolating polynomial for a sequence,  $S$ , defined recursively using a generating polynomial,  $g$ , of finite degree  $n$  and a known  $(x_0, y_0)$  pair as follows,

With  $(x_0, y_0)$  and  $h$  and  $g(x)$  given, the recursive formula defines an infinite set  $S$ ,

$$S = \{(x_i, y_i)\}_{i=-\infty}^{\infty}$$
$$\begin{cases} x_{i+1} = x_i + h, \\ y_{i+1} = y_i + g(x_{i+1}) \end{cases}$$

**Theorem: There exists a unique finite degree polynomial interpolant,  $p$ , of the infinite set  $S$  generated by the recursive formula above. If the generator polynomial  $g \in \mathbb{P}_n$  then the interpolant  $P$  is of degree  $n + 1$ .**

$$p(x_i) = y_i, \quad \text{for } \forall x_i, y_i \in S$$

If the following two conditions are met, then  $p$  interpolates  $S$ ,

$$p(x_0) = y_0 \quad \text{(condition 1)}$$

$$p(x_i + h) = p(x_i) + g(x_i + h), \quad \forall x_i, y_i \in S \quad \text{(condition 2)}$$

$$h \neq 0$$

## Existence

$$p(x_i + h) = p(x_i) + g(x_i + h) \quad (2)$$

Using vector representations of  $p$ ,  $g$ , and  $x$  and an  $n + 1 \times n + 1$  matrix,  $D_h$ , for  $p(x + h)$ 's representation,

$$\vec{p} = \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix}, \vec{g} = \begin{pmatrix} b_0 \\ b_1 \\ \vdots \\ b_n \end{pmatrix}, \vec{x}^T = \begin{pmatrix} x^0 \\ x^1 \\ \vdots \\ x^n \end{pmatrix}$$

$$D_h = \begin{pmatrix} d_{00} & d_{01} & \dots & d_{0n} \\ 0 & d_{11} & \dots & d_{1n} \\ \vdots & \vdots & d_{ij} & \vdots \\ 0 & 0 & \dots & d_{nn} \end{pmatrix}, d_{ij} = \binom{j}{i} (-h)^{j-i}$$

Then equation (2) can be represented,

$$\vec{x} D_h \vec{p} = \vec{x} \vec{p} + \vec{x} D_h \vec{g} \quad (2.2)$$

Rearranging equation (2.2),

$$D_h^{-1}(D_h - I)\vec{p} = \vec{g}$$

The properties of  $d_{ij}$  make the diagonal elements of  $D_h$  all equal to one,

$$d_{ij} = \binom{j}{i}(-h)^{j-i} = 1 \text{ if } j = i$$

When the identity matrix is subtracted from  $D_h$ , the matrix remains upper triangular but with zero elements along its diagonal,

Let,

$$A = D_h - I$$

$$A = \begin{pmatrix} 0 & d_{01} & \dots & d_{0n} \\ 0 & 0 & \dots & d_{1n} \\ \vdots & \vdots & 0 & \vdots \\ 0 & 0 & \dots & 0 \end{pmatrix}$$

This leads to an inconsistent system for equation (2.2) if  $\vec{g}$  has a non zero number as its final entry. Therefore if  $p \in \mathbb{P}_n$  then  $g \in \mathbb{P}_{n-1}$  for a solution to exist.

$$\begin{pmatrix} 0 & d_{01} & \dots & d_{0n} \\ 0 & 0 & \dots & d_{1n} \\ \vdots & \vdots & 0 & \vdots \\ 0 & 0 & \dots & 0 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} b_0 \\ b_1 \\ \vdots \\ 0 \end{pmatrix} \quad (2.2)$$

Additionally, due to the other properties of  $d_{ij}$ , if  $h \neq 0$ , then every element of  $A$  above its main diagonal will  $\neq 0$ , leading to a pivot column in all other rows. Additionally,  $g(x + h)$  can be written as another polynomial of the same degree so  $D_h^{-1}$  can temporarily be ignored.

We now have a system of  $n$  equations that can be used to solve for the  $n$  unknowns  $a_1, a_2, \dots, a_n$ .

$$\begin{pmatrix} d_{01} & \dots & d_{0n-1} \\ 0 & d_{12} & \dots & d_{1n-1} \\ \vdots & & \vdots & \\ 0 & \dots & d_{n-1n-1} \end{pmatrix} \begin{pmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix} \quad (2.3)$$

Therefore,

$$\forall g \in \mathbb{P}_{n-1}, \exists p \in \mathbb{P}_n \mid p(x + h) = p(x) + g(x + h) \quad (\text{existence})$$

## Uniqueness

For any generating polynomial,  $g \in \mathbb{P}_n$ , for the a sequence  $S$ , there exists a unique polynomial,  $p$ , that satisfies

$$p(x_0) = y_0 \quad (\text{condition 1})$$

$$p(x + h) = p(x) + g(x + h), \quad h \neq 0 \quad (\text{condition 2})$$

The unique polynomial,  $p$ , is of finite degree  $n + 1$ .

Above it was shown that a polynomial of finite degree  $n + 1$  exists that satisfies conditions (1) and (2), (taking  $n + 1$  to be the smallest degree where the system is not inconsistent,  $n + 1 = (\deg(g) + 1)$ ).

An infinite set of points,  $S$ , that  $p$  must interpolate can be generated as follows,

$$\begin{aligned} &\text{Given } (x_0, y_0), \\ &y_{i+1} = y_i + g(x_i), \\ &x_{i+1} = x_i + h, \\ &S = \{(x_i, y_i)\}_{i=0}^{\infty} \end{aligned}$$

This method gives an infinite set of points,  $S$ , all distinct. Assume another polynomial,  $r$ , of degree  $m \geq n$  interpolates all the points in  $S$ . Then by subtracting  $p$  and  $r$  we get a third polynomial,  $s$ , of maximum degree  $m$ ,

$$s = p - r, \quad s \in \mathbb{P}_m$$

Because  $p$  and  $r$  both interpolate each point in the infinite set  $S$ , the polynomial,  $s$  will have an infinite amount of zeros.

$$s(x_i) = 0, \quad \forall x_i \in S$$

Because  $m$  is finite,  $S$  must be the zero polynomial. Therefore,

$$p(x) = r(x)$$

And  $p$  is unique and of degree  $n + 1$ .

## **solving for $p$**



The coefficients,  $a_1, a_2, \dots, a_n$  can be solved for using the system outlined in (2.3). With  $n$  unknowns and  $n$  equations. The unique solution that satisfies conditions (1) and (2) can be found by using equation (1) to solve for  $a_0$ .

$$p(x_0) = y_0 \quad (1)$$

This shows that a unique solution exists and can be solved for, for each system with  $p \in \mathbb{P}_k$  where  $k \geq n$ .

$$\forall g \in \mathbb{P}_n, \quad \exists p \in \mathbb{P}_{n+1}, \text{ where } h \neq 0 \quad \text{s.t.},$$

$$p(x_0) = y_0 \quad (1)$$

$$p(x + h) = p(x) + g(x + h) \quad (2)$$

## Connection to Power Sums

let,

$$\sum_{k=1}^n k^d = 1^d + 2^d + \dots + n^d$$

define a " $d$ th" order power sum of the first  $n$  integers.

The above method can be used for finding an interpolating polynomial,  $p_d \in \mathbb{P}_{d+1}$ , of the  $d$ th order power sum, for  $\forall d \in \mathbb{N}$ , such that,

$$p_d(n) = \sum_{k=1}^n k^d, \quad \forall n \in \mathbb{N}$$

The following property holds for any  $d$ th order power sum,

$$\sum_{k=1}^{n+1} k^d = \sum_{k=1}^n k^d + (n+1)^d$$

By induction the following two properties are sufficient conditions for a polynomial,  $p$ , to interpolate a  $d$ th order power sum,

$$p_d(1) = \sum_{k=1}^1 k^d, \quad (c1)$$

$$p_d(n+1) = p_d(n) + (n+1)^d, \text{ for all } n \in \mathbb{N} \quad (c2)$$

## CODE

In [141]:

```
def power_sum( n, d=1 ):
    """ power_sum( n ) - power_sum( n-1 ) = n**d """

    if np.isscalar(n):
        # must be an integer
        n = int(n)
        s = 0
        for i in range(1,n+1):
            s += i**d
        return s

    return np.array([ power_sum( ni, d=d ) for ni in n ])
```

In [142]:

```
#
# create Horizontal Shift Matrix
#

def poly_shift_matrix(order, h):
    # h -- horizontal shift amount

    # n+1 coeficients to solve for -- n+1 x n+1 shift matrix
    n = order + 1

    # pascal matrix
    P = pascal(n, kind='upper')

    # create H
    H = np.eye(n)
    _h = 1
    for k in range(n):
        for i in range(n-k):
            H[i,i+k] = _h
        _h *= h

    # create shift matrix - flip for numpy convention
    S = np.flip(H * P)

    # return
    return S

# recursion:  $p(x+h) = p(x) + g(x+h)$ ,  $p(x_0) = y_0$ 
def interpolate_recursion(g, x0, y0, h=1):
    # g is a numpy polynomial

    # get degree of p, n, from degree of g, n-1
```

```

n = g.order + 1

# get shift matrix
S = poly_shift_matrix(n, h)

# Find  $A = S^{-1}(S - I)$ , system  $Ap = q$  -- solve for p after removing row and column
# corresponding to the constant term

I = np.eye(n+1)
A = np.linalg.inv(S) @ (S - I)

# remove first row and last column
A = A[1:,:-1]

p_pad = np.zeros(shape=(n+1,))
p_pad[:-1] = np.linalg.solve(A,g)

p_raw = np.polyld(p_pad)

# solve system for final coefficient
diff = p_raw(x0)

if x0 != 0:
    p_raw += (y0 - diff)/x0

else:
    p_raw += y0

return np.polyld(p_raw)

#
# settings
#

g_degree = random.randint(1,30)

x0 = random.randint(0,100)
y0 = power_sum(x0, d=g_degree)

g = np.zeros(shape=(g_degree+1,))
g[0] = 1
g = np.polyld(g)

# find interpolants
p = interpolate_recursion(g, x0, y0 )

#
# display
#
import matplotlib.pyplot as plt
import pandas as pd
import seaborn

```

```
N = 25
```

```
# predicted
```

```
x_values = np.linspace(0,N,N*2 + 1)
```

```
y_hat_values = p(x_values)
```

```
# true
```

```
int_x_values = np.array(list(range(1,N)))
```

```
y_values = power_sum(int_x_values, d=g_degree)
```

```
df = pd.DataFrame(list(zip(x_values, y_hat_values)), columns=['x', 'interp'])
```

```
df = df.set_index('x')
```

```
seaborn.set(style='ticks')
```

```
fig, ax = plt.subplots()
```

```
ax.grid(True)
```

```
ax = seaborn.lineplot(data=df, color="coral")
```

```
ax = seaborn.scatterplot(x=int_x_values, y=y_values, color="red")
```

```
plt.show()
```

