Computational Graphics: Lecture 6

The CVDlab Team

Thu, Mar 13, 2014

Outline: Algebra3

1. Introduction to `pyplasm`
2. Matrices

Introduction to pyplasm

PLaSM Basics

PLaSM = Geometric extension of the FL language by Backus (developed at IBM Research)

> *A. Paoluzzi, V. Pascucci and M. Vicentino: Geometric Programming: A Programming Approach to Geometric Design. ACM Transactions on Graphics 14(3): 266-306 (1995)*

1. geometric calculus in FL-style

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1. geometric calculus in FL-style
2. dimension independence
3. dynamic typing
4. higher-level operators
5. arity: always 1 (number of arguments of functions)
6. small set of predefined functionals
7. names of functions: all-caps

PLaSM Basics (AA: Apply-to-All)

```
>>> AA(SUM)([[1,2,3],[4,5,6]])
[6,15]
```

PLaSM Basics (AA: Apply-to-All)

```
>>> AA(SUM)([[1,2,3],[4,5,6]])
[6,15]

>>> mat = [[1,2,3],[4,5,6]]
>>> [sum(v) for v in mat]
[6,15]
```

PLaSM Basics (DISTL: DISTribute-Left)

```
>>> DISTL([2,[1,2,3]])
[[2,1],[2,2],[2,3]]
```

PLaSM Basics (DISTL: DISTribute-Left)

```
>>> DISTL([2,[1,2,3]])
[[2,1],[2,2],[2,3]]

>>> DISTL([2,[]])
[]
```

PLaSM Basics (TRANS: TRANSpose)

```
>>> TRANS([[1,2,3],[10,20,30],[100,200,300]])
[[1,10,100],[2,20,200],[3,30,300]]
```

PLaSM Basics (TRANS: TRANSpose)

```
>>> TRANS([[1,2,3],[10,20,30],[100,200,300]])
[[1,10,100],[2,20,200],[3,30,300]]

>>> TRANS([[1,2,3,4,5],[10,20,30,40,50]])
[[1,10],[2,20],[3,30],[4,40],[5,50]]
```

PLaSM Basics (TRANS: TRANSpose)

```
>>> TRANS([[1,2,3],[10,20,30],[100,200,300]])
[[1,10,100],[2,20,200],[3,30,300]]

>>> TRANS([[1,2,3,4,5],[10,20,30,40,50]])
[[1,10],[2,20],[3,30],[4,40],[5,50]]

>>> TRANS([[],[]])
[]
```

PLaSM Basics (arithmetic ops)

```
>>> PROD([3,4])
12
```

PLaSM Basics (arithmetic ops)

```
>>> PROD([3,4])
12

>>> PROD([[1,2,3],[4,5,6]])
32.0
```

PLaSM Basics (arithmetic ops)

```
>>> PROD([3,4])
12

>>> PROD([[1,2,3],[4,5,6]])
32.0

>>> SUM([3,4])
7
```

PLaSM Basics (arithmetic ops)

```
>>> PROD([3,4])
12

>>> PROD([[1,2,3],[4,5,6]])
32.0

>>> SUM([3,4])
7

>>> SUM([[1,2,3],[4,5,6]])
[5, 7, 9]
```

PLaSM Basics (product scalar by vector)

```
>>> SCALARVECTPROD([3,[1,2,3]])
[3, 6, 9]
```

PLaSM Basics (product scalar by vector)

```
>>> SCALARVECTPROD([3,[1,2,3]])
[3, 6, 9]

>>> SCALARVECTPROD([4,[10,20,30]])
[40, 80, 120]
```

Pyplasm: Exercise 1 (INNERPROD)

The inner (or scalar) product of $a, b \in \mathbb{R}^m$ is a number

$$\mathtt{INNERPROD} : \mathbb{R}^m \times \mathbb{R}^m \to \mathbb{R} : (u, v) \mapsto \sum_{i=1}^{m} \mathbf{u}_i \mathbf{v}_i$$

Pyplasm: Exercise 1 (INNERPROD)

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```
>>> u = [1,2,3]
>>> v = [10,20,30]
>>> INNERPROD([u, v])
140
```

Pyplasm: Exercise 2 (VECTNORM)

The norm of a vector $a \in \mathbb{R}^m$ is a number.

$$\texttt{VECTNORM} : \mathbb{R}^m \to \mathbb{R} : v \mapsto \sqrt{\sum_{i=1}^{m} \mathbf{v}_i^2}$$

Pyplasm: Exercise 2 (VECTNORM)

The norm of a vector $a \in \mathbb{R}^m$ is a number.

$$\mathtt{VECTNORM} : \mathbb{R}^m \to \mathbb{R} : v \mapsto \sqrt{\sum_{i=1}^{m} \mathbf{v}_i^2}$$

```
>>> a = [1,2,3]
>>> VECTNORM(a)
3.7416573867739413
```

Pyplasm: Exercise 3 (UNITVECT)

The unit vector is a function

$$\texttt{UNITVECT} : \mathbb{R}^m \rightarrow \mathbb{R}^m : v \mapsto \frac{v}{|v|}$$

Pyplasm: Exercise 3 (UNITVECT)

The unit vector is a function

$$\mathtt{UNITVECT} : \mathbb{R}^m \to \mathbb{R}^m : v \mapsto \frac{v}{|v|}$$

```
>>> v = [1,2,3]
>>> UNITVECT(v)
[0.2672612369060516 4, 0.5345224738121033, 0.8017836809158325]
```

Pyplasm: Exercise 3 (UNITVECT)

The unit vector is a function

$$\mathtt{UNITVECT} : \mathbb{R}^m \to \mathbb{R}^m : v \mapsto \frac{v}{|v|}$$

```
>>> v = [1,2,3]
>>> UNITVECT(v)
[0.26726123690605164, 0.5345224738121033, 0.8017836809158325]

>>> VECTNORM(UNITVECT(v))
0.9999999403953552   1
```

Pyplasm: Exercise 4 (SUM)

SUM adds m vectors in $\mathbb{R}^n$, i.e. the rows of a matrix in $\mathbb{R}^m_n$:

```
>>> a = [1,2,3]
>>> a
[1, 2, 3]

>>> b = [10,20,30]
>>> b
[10, 20, 30]

>>> SUM([a,b])
[11, 22, 33]
```

Pyplasm: Exercise 5 (SUM)

```
>>> a = range(10)
>>> a
[0, 1, 2, 3, 4, 5, 6, 7, 8, 9]

>>> b =  [10*k for k in range(10)]
>>> b
[0, 10, 20, 30, 40, 50, 60, 70, 80, 90]

>>> SUM([a,b])
[0, 11, 22, 33, 44, 55, 66, 77, 88, 99]

>>> c =  [100*k for k in range(10)]
>>> c
[0, 100, 200, 300, 400, 500, 600, 700, 800, 900]

>>> SUM([a,b,c])
[0, 111, 222, 333, 444, 555, 666, 777, 888, 999]
```

Pyplasm: Exercise 6 (MATSUM)

Assignment

Write a function that adds any two matrices $[A], [B]$ (compatible by sum).
Both $[A], [B]$ must belong to the same linear space $\mathbb{R}^m_n$

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```
>>> def MATSUM(args):
...     return  AA(AA(SUM)) (AA(TRANS)(TRANS(args)))

>>> A = [ [1,2,3], [4,5,6], [7,8,9] ]
>>> B = [ [10,20,30], [40,50,60], [70,80,90] ]
```

Pyplasm: Exercise 6 (MATSUM)

Assignment

Write a function that adds any two matrices $[A], [B]$ (compatible by sum).
Both $[A], [B]$ must belong to the same linear space $\mathbb{R}_n^m$

```
>>> def MATSUM(args):
...     return  AA(AA(SUM)) (AA(TRANS)(TRANS(args)))

>>> A = [ [1,2,3], [4,5,6], [7,8,9] ]
>>> B = [ [10,20,30], [40,50,60], [70,80,90] ]

>>> MATSUM([A,B])
[ [11,22,33], [44,55,66], [77,88,99] ]

>>> MATSUM([A,B,A])
[ [12,24,36], [48,60,72], [84,96,108] ]

>>> MATSUM([A,B,B,A])
[ [22,44,66], [88,110,132], [154,176,198] ]
```

Pyplasm: Exercise 7 (MATPROD)

Write a function that multiplies two matrices (compatible by product)

Remember that

$$A \in \mathbb{R}^m_n, \quad B \in \mathbb{R}^n_p, \quad \text{and} \quad C = AB \in \mathbb{R}^m_p,$$

with

$$C = \left(\begin{array}{c} c^i_j \end{array} \right) = \left(\begin{array}{c} \mathbf{A^i B_j} \end{array} \right), \qquad 1 \leq i \leq m, 1 \leq j \leq p,$$

where $\mathbf{A^i}$ is the *i*-th row of **A**, and $\mathbf{B_j}$ is the *j*-th column of **B**.

Pyplasm: Exercise 7 (MATPROD) – Solution

Write a function that multiplies two compatible matrices

```
>>> def MATPROD(args):
...     A,B = args
...     return AA(AA(INNERPROD)) (AA(DISTL) (DISTR ([A, TRANS (B)])))
```

Pyplasm: Exercise 7 (MATPROD) – Solution

Write a function that multiplies two compatible matrices

```
>>> def MATPROD(args):
...     A,B = args
...     return AA(AA(INNERPROD)) (AA(DISTL) (DISTR ([A, TRANS (B)])))

>>> A = [[1,2,3],[4,5,6],[7,8,9]]
>>> B = [[1,2,3],[4,5,6],[7,8,9]]
>>> MATPROD ([A,B])
[ [30, 36, 42], [66,81,96], [102,126,150] ]
```

Pyplasm: Exercise 7 (MATPROD) – Solution

Write a function that multiplies two compatible matrices

```
>>> def MATPROD(args):
...     A,B = args
...     return AA(AA(INNERPROD)) (AA(DISTL) (DISTR ([A, TRANS (B)])))

>>> A = [[1,2,3],[4,5,6],[7,8,9]]
>>> B = [[1,2,3],[4,5,6],[7,8,9]]
>>> MATPROD ([A,B])
[ [30, 36, 42], [66,81,96], [102,126,150] ]

>>> C = [[1,2,3],[4,5,6]]
>>> D = [[1,2],[4,5],[7,8]]
>>> MATPROD ([C,D])
[ [30,36], [66,81] ]
```

Pyplasm: Exercise 8 (some array operators)

Look at some PLaSM operators on arrays

```
>>> N(3) (0)   # REPEAT
[0,0,0]
>>> N(3) ([0,1])
[ [0,1], [0,1], [0,1] ]
```

Pyplasm: Exercise 8 (some array operators)

Look at some PLaSM operators on arrays

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>>> N(3) (0)    # REPEAT
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>>> N(3) ([0,1])
[ [0,1], [0,1], [0,1] ]

>>> NN(3) ([0,1])  # REPeat LISt & CAtenate -- REPLICA
[ 0,1, 0,1, 0,1 ]
```

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Look at some PLaSM operators on arrays

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>>> N(3) (0)    # REPEAT
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>>> AR ([ [0,0,0], 1 ])  # Append Rigth
[0,0,0,1]
```

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>>> NN(3) ([0,1])  # REPeat LIst & CAtenate -- REPLICA
[ 0,1, 0,1, 0,1 ]

>>> AR ([ [0,0,0], 1 ])  # Append Rigth
[0,0,0,1]

>>> AL ([ 1, [0,0,0] ]) # Append Left
[1,0,0,0]
```

Pyplasm: Exercise 9 (VECTPROD)

the vector product **w** of vectors in $\mathbb{R}^3$ id defined as the function

$$\mathbb{R}^3 \times \mathbb{R}^3 \to \mathbb{R}^3 : (\mathbf{u}, \mathbf{v}) \mapsto \det \begin{pmatrix} \mathbf{e}_0 & \mathbf{e}_1 & \mathbf{e}_2 \\ u_x & u_y & u_z \\ v_x & v_y & v_z \end{pmatrix}$$

Therefore we can write, for the vector product of two 3D vector:

```
>>> def VECTPROD(args):
...     u,v = args
...     w = [0,0,0]
...     w[0] = u[1]*v[2] - u[2]*v[1]
...     w[1] = u[2]*v[0] - u[0]*v[2]
...     w[2] = u[0]*v[1] - u[1]*v[0]
...     return w
```

Pyplasm: Exercise 9 (VECTPROD)

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...     w[0] = u[1]*v[2] - u[2]*v[1]
...     w[1] = u[2]*v[0] - u[0]*v[2]
...     w[2] = u[0]*v[1] - u[1]*v[0]
...     return w

>>> VECTPROD([[1,0,0], [0,1,0]])
[0,0,1]
>>> VECTPROD([[1,1,0], [0,1,0]])
[0,0,1]
```

Pyplasm: Exercise 10

```
>>> from random import random

>>> def randomPoints(m, sx=1, sy=1):
...     def point():
...         return [random() * sx, random() * sy]
...     return [point() for k in range(m)]
```

Pyplasm: Exercise 10

```
>>> from random import random

>>> def randomPoints(m, sx=1, sy=1):
...     def point():
...         return [random() * sx, random() * sy]
...     return [point() for k in range(m)]

>>> verts = randomPoints(200, 2*PI, 2)
>>> obj = MKPOL([verts, AA(LIST)(range(200)), None])
```

Pyplasm: Exercise 10

```
>>> from random import random

>>> def randomPoints(m, sx=1, sy=1):
...     def point():
...         return [random() * sx, random() * sy]
...     return [point() for k in range(m)]

>>> verts = randomPoints(200, 2*PI, 2)
>>> obj = MKPOL([verts, AA(LIST)(range(200)), None])

>>> VIEW(obj)
```

Pyplasm: Exercise 11

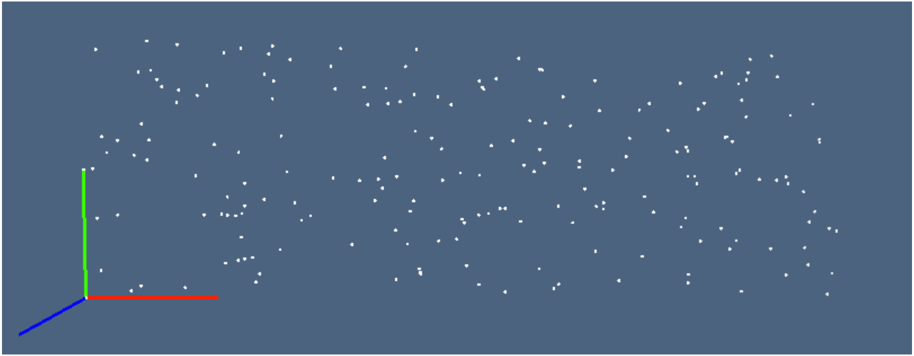

Figure 1: 200 random points in $[0, 2\pi] \times [0, 2] \subset \mathbb{E}^2$

Pyplasm: Exercise 12

coordinate functions

```
>>> def x (p):
...     u,v = p
...     return v * COS(u)

>>> def y (p):
...     u,v = p
...     return v * SIN(u)
```

Pyplasm: Exercise 12

coordinate functions

```
>>> def x (p):
...     u,v = p
...     return v * COS(u)

>>> def y (p):
...     u,v = p
...     return v * SIN(u)

>>> obj = MAP([ x,y ])(obj)
```

Pyplasm: Exercise 12

coordinate functions

```
>>> def x (p):
...       u,v = p
...       return v * COS(u)

>>> def y (p):
...       u,v = p
...       return v * SIN(u)

>>> obj = MAP([ x,y ])(obj)

>>> VIEW(obj)
```

Pyplasm: Exercise 12 (4/4)

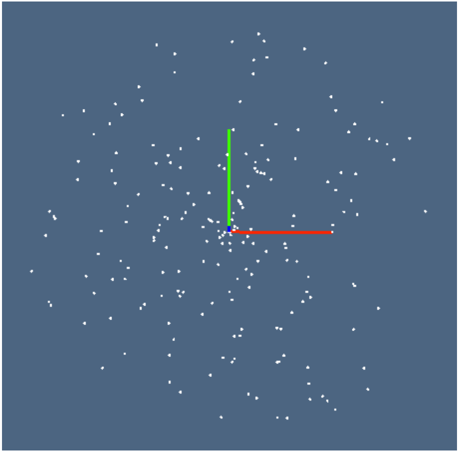

Figure 2: 200 random points within the 2D "ball" of radius 2

From PLaSM to Pyplasm

application (binary infix operator :) to (...)

$$f : x \rightarrow f(x)$$

composition (binary infix operator) to COMP

$$f \sim g \rightarrow COMP([f, g])$$

construction (of a vector function) to CONS

$$[f, g] : x \rightarrow CONS([f, g])(x)$$

sequence (arrow parentheses < ... >) to list

$$f : \langle x_1, x_2, \ldots, x_n \rangle \rightarrow f([x_1, x_2, \ldots, x_n])$$

From PLaSM to Pyplasm

The original FL syntax

```
hpc = MAP:f:dom
WHERE
    f = [COS~S1, SIN~S1],
    dom = INTERVALS: (2*PI): 24
END;

DRAW:hpc
```

Ported syntactically to python

```
>>> f = CONS([ COMP([COS,S1]), COMP([SIN,S1]) ])
>>> dom = INTERVALS(2*PI)(24)
>>> hpc = MAP(f)(dom)
>>> VIEW(hpc)
```

Using properly the Python syntax

The function to be mapped is from *d*-points to lists of coordinate functions $\mathbb{R}^d \to \mathbb{R}$

```
>>> def circle(p):
...     alpha = p[0]
...     return [COS(alpha), SIN(alpha)]
```

Using properly the Python syntax

The function to be mapped is from *d*-points to lists of coordinate functions $\mathbb{R}^d \to \mathbb{R}$

```
>>> def circle(p):
...     alpha = p[0]
...     return [COS(alpha), SIN(alpha)]

>>> obj = MAP(circle)(INTERVALS(2*PI)(32))
>>> VIEW(obj)
```

In case of a curve, $d = 1$

Current plasm.js Library

a subset of pyplasm: see fenvs.py)

AA
AL
APPLY
AR
BIGGER
BIGGEST
BOUNDARY
BUTLAST
CART
CAT
CENTROID
CIRCLE
CLONE
CODE
COMP
CONS
CUBE
CUBOID
CYLSOLID
CYLSURFACE
DISK
DISTL
DISTR
DIV
EMBED
EXPLODE
EXTRUDE
FIRST
FREE
Graph
GRAPH
HELIX
ID
IDNT
IDNT
INNERPROD
INSL
INSR
INTERVALS
INV
ISFUN
ISNUM
K
LAST
LEN
LINSPACE1D
LINSPACE2D
LINSPACE3D
LIST
MAP
MAT
MATPROD
MATSUM
MUL
PointSet
POLYLINE
POLYMARKER
PRECISION
PRINT
PROD
PROGRESSIVE_SUM
QUADMESH
R
REPEAT
REPLICA
REVERSE
S
S0
S1
S2
S3
S4
SET
SIMPLEX
SIMPLEXGRID
SimplicialComplex
SKELETON
SMALLER
SMALLEST
SORTED
SUB
SUM
T
TAIL
Topology
TORUSSOLID
TORUSSURFACE
TRANS
TREE
TRIANGLEARRAY
TRIANGLEFAN
TRIANGLESTRIP
UNITVECT
VECTNORM
VECTPROD

Matrices

Efficient matrix calculus with `NumPy` and `SciPy`

docs.scipy.org

NumPy for Matlab Users

‘array’ or ‘matrix’? Which should I use?

Use arrays

- They are the standard vector/matrix/tensor type of numpy. Many numpy function return arrays, not matrices.

The only disadvantage of using the array type is that you will have to use dot instead of * to multiply (reduce) two tensors (scalar product, matrix vector multiplication etc.).

‘array’ or ‘matrix’? Which should I use?

Use arrays

- They are the standard vector/matrix/tensor type of numpy. Many numpy function return arrays, not matrices.
- There is a clear distinction between element-wise operations and linear algebra operations.

The only disadvantage of using the array type is that you will have to use dot instead of * to multiply (reduce) two tensors (scalar product, matrix vector multiplication etc.).

‘array’ or ‘matrix’? Which should I use?

Use arrays

- They are the standard vector/matrix/tensor type of numpy. Many numpy function return arrays, not matrices.
- There is a clear distinction between element-wise operations and linear algebra operations.
- You can have standard vectors or row/column vectors if you like.

The only disadvantage of using the array type is that you will have to use dot instead of * to multiply (reduce) two tensors (scalar product, matrix vector multiplication etc.).

Assignments

READ:

NumPy: The Numpy array object

References

Python Scientific Lecture Notes