Computational Medicine:

Oberon-based Simulation, Control, and Signal processing

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Computers in Medicine





Image acquisition Image data flow Image analysis Image retrieval



Biosianals

-Signal processing

-Signal analysis

Biosignals for machine contro

The feedback loop of medicine







Digital Signal Processing in Oberon: DSP/Wavelets Libraries



Hunziker, Jansen, Conen, unpublished





Matrix Libraries

Based on MatrixOberon language (ETHZ) SIMD accelerated – multicore capable



Linear Algebra Library (Uni Basel) MatrixBase

MatrixTransforms (Symmetric,Skew, Householder, ...) MatrixStandardSolvers (LU,Gauss-Jordan,QR,Cholesky) MatrixIterativeSolvers (Gauss-Seidel,Jacobi,SOR,CG) MatrixMultigridSolvers (AMG) MatrixKrylovSolvers (GMRES, GMRES(k), MINRES, ..) MatrixPreconditioners (SPAI) MatrixOptim (n-dim function minimization) MatrixNeuralNetwork

Matrix Libraries - example

Pseudocode (Wikipedia)

Choose an initial guess φ^0 to the solution for k := 1 step 1 until convergence do for i := 1 step until n do $\sigma = 0$ for j := 1 step until n do if j != i then $\sigma = \sigma + a_{ij}\phi_j^{(k-1)}$

end if end (j-loop)

 $\phi_i^{(k)} = \frac{(b_i - \sigma)}{a_{ii}}$

end (i-loop) check if convergence is reached

end (k-loop)

Implementation (incl SIMD)

NODULE //atrixDemolacobi; NYPE Mattix = ARRAY [..., ..] OF LONGREAL; Vector = ARRAY [...] OF LONGREAL;

PROCEDURE Diag (CONST A: Matrix): Vector

PROCEDURE Inv (CONST V: Vector): Vector;

(## Solve Ax=b_using Jacobi Algorithm. #)
PROCEDURE Jacobi*(CONST_A: Matrix: COI

PROCEDURE Iacobi+(CONST A: Matrix; CONST b: Vector; VAR x: Vector; threshold: LONGREAL) VAR invd, d, r: Vector; BFGIN

d := Diag(A); invd := Inv(d);

LOOP r := A + s: (* matrix-vector product. *

IF AathLsqt((r - b) ++ (r - b)) < threshold THEN RETURN x END; (* '+*' is inner product

x := (b - r + d .* x) .* invd; (* '.* ' is elementwise product *) END:

END Jacobi;

END MatrixDemolacob

The need for Tensor(Multilinear) Tools Nonregular sampled data: an important problem in medical data analysis

- Patient monitoring: irregular measurement intervals
- Computer tomography:

circular data acquisition, divergent X-ray beam

4D Ultrasound:

"random" data from 3D space and time => resample to 3D moving dataset in regular grid

Simple example:

Reconstruct map of Switzerland from arbitrary samples efficiently, least square optimal, with minimal curvature

10 cities 7 mountains



Solution:

-*Cubic B-splines:* smooth interpolants with shortest support (->speed) and minimal curvature

- regularized (tolerate missing data) least square problem

Real world example

- [3D+t] data acquisition from beating heart 1 second =
 - 128*128*32*32 =10⁷ Samples at arbitrary location
 - $128*128*32*32 = 10^7$ cartesian grid points
 - => sparse linear system size 10⁷*10⁷ 10⁷ unknowns complicated sparsity pattern
 A high performance computation problem !
 - Suited for Oberon ?

The solution: multilinear (tensor) algebra in 'high performance Oberon'

Analysis of the problem:

- Original data share spatial connectivity which is
 +/- lost in matrix representation (sparsity)
- This least squares problem S^TSx=S^Tb is actually a tensor problem with



Tensor Oberon

- The language extension (ETHZ)
- The tensor algebra library
 - Data structures
 - Multilinear signal processing
 - Multilinear solvers



Grid/Cluster computation with Tensor Oberon









The Basel Telemedicine Concept:



Patient A.B. Prähospital EKG





Structure of the bronchial tree in man

Several morphometric models: Symmetric <-> asymmetric <-> monopodial Connected <-> not connected

Model of Weibel, 1963 Symmetric, connected most widely used

Model of Horsefield, 1971: Asymmetric, connected

Model of Yeh, 1979: Symmetric, not connected

Model of Ross 1957: asymmetric, not connected



1 2 3 4 5 6 7 8 9 Strahler Order

Background of lung simulation

Clinical studies with new strategies to treat patients with lung disease, but we realized that it is difficult to understand what is going on in the patient

Simple mechanical lung models. *disappointing* yielding very different results from what is observed in patients.

Clinical studies : conflicting results:

- improvement in volumes, pressures
- few effects on blood gases

Key pathophysiological characteristics of airway disease:

- narrowing of airways,
 - predominantly distal (COPD)
 - or proximal (larynx edema, obstruction; Asthma)
- Inhomogenous pathology in different parts of lung
- Laminar versus turbulent flow conditions
- Air trapping (dynamic over-inflation)
- Dynamic airway collapse
- Secretions

complex physics ! Is simulation feasible ?

Fluid dynamics of air flow

Laminar flow conditions: Hagen-Poiseuille Equation

pressure loss depends on airway radius, length, gas viscosity; minor influence of surface roughness

Turbulent flow conditions: pressure loss depends on airway radius, length, gas density strong dependence of surface roughness

Probability of laminar versus turbulent flow: Depends on Re (Reynold's number), which is made up of Gas Velocity, Gas Density, Diameter, XX

Pressure loss at bifurcations Depends on bifurcation shape, angle, XX





The system

-based on COSIMO **Oberon package** -elements of circulation system modelled as connected blocks

Damne

r 10 +

Lung

0

A 4-1-4

In #

0

Flow



airway

pressure

3

airway

gradient

1

The normal lung



The small endotracheal tube/ tracheal obstruction

airway

gradient 2

alveolar

pressure

4



The combination of lung disease (COPD) and fear (rapid breathing) leads to dangerous dynamic hyper-inflation of the lung



Conclusions

Real-time simulation of lung disease based on computational fluid dynamics is feasible

yields ventilation patterns that correspond to those observed in real patients with well-defined airway disease

allows quantitative assessment of new ventilation strategies

resulted in new insights in effects of Helium ventilation in COPD patients

will allow training in ventilation management of patients with airway disease on the ICU without danger for the patients



Catheter-based pumps in shock:

Prediction of efficacy by computational disease modeling and clinical observations

Cardiogenic Shock: a highly lethal syndrome



=> Do everything to give them this chance

Physiology of shock in AMI

Healthy myocardium: can be stimulated

Necrotic myocardium: no recovery. High wall stress stretches necrosis = early dilatation in AMI

Stunned myocardium: able to recover, but late (hours to days)

Ischemic myocardium in border zone: able to improve or to worsen rapidly

Myocardium in remote territories with CAD: potential for ischemia

Shock physiology in AMI is complex

Understanding is key to rational treatment.



percutaneous mechanical support







Catheter based axial pumps – transaortic approach Nimbus Hemopump Impella Recover F14. Ext motor, up to 1.5-2L/min. Urban et F12 (2.5L/'), F21 (5L/')

The Basel Heart Simulator

Computational fluid dynamics simulation of the cardiovascular system



A system of differential equations for mathematical modeling of multiple segments of the circulation based on in vivo- data.

Full biomechanical modellina: heart, arteries, peripheral circulation & venous system

> Impact of systolic & diastolic abnormalities, ischemia, drug effects & can be studied.

Hunziker & al, 2006

Suehling, Hunziker, Circulation 2005

Modeling: Numerical Integration

BEGIN Integrator.Dynamic := Dynamic; Integrator.Integ := IntegDP54.DormandPrince54; horizon:=0.3;

PROCEDURE In

tegrator.InitLists:	
10° EUXily 2021-Ab0/075; V301-Ab0/075; gmis.Dec15tate5igmi(V1, 11, V10, horiz gmis.Dec15tate5igmi(V2, 12, V20, horiz gmis.Dec15tate5igmi(V3, 13, V30, horiz gmis.Dec15tate5igmi(V5, 15, V30, 0); gmis.Dec15tate5igmi(V5, 15, V30, 0); gmis.Dec15tate5igmi(V5, 16, V40, 0); gmis.Dec15tate5igmi(V6, 19, V90, 0); gmis.Dec15tate5igmi(V7, 17, V70, 0); gmis.Dec15tate5igmi(V7, 19, V90, 0); gmis.Dec15tate5igmi(V7, 19, V90, 0); gmis.Dec15tate5igmi(V7, 19, V90, 0); gmis.Dec15tate5igmi(V7, 19, V20, 0); gmis.Dec15tate5igmi(V7, V70, 0); gmis.Dec15tate5igmi(V7, V70, 0); gmis.Dec15tate5igmi(V7, V70, V70, 0); gmis.Dec15tate5igmi(V7, V70, V70, 0); gmis.Dec15tate5igmi(V7, V70, V70, 0); gmis.Dec15tate5igmi(V70,	<pre>(Presidences) Fit = AVSate(P, P2, OldAntic/slveState); (eIF (oldAntic/slveState-AVRD, 8 (R1=V/PC) THEN LVESX=Signals.interpolateSignal(s/1,Integrator.time=0.01);END; (+aortic-valve has just closed+)=) (FV-LVESY=THEN LVESX=V-I END; oldAntic/slveState=R1; (+aortic-valve resistance, a function of valve opening+) R2:= AoR; (+aortic-valve resistance); R3:= AoR; (+aortic-valve resistance); R4:= PenpherBisstanceVasconstriction/Vasodilator); (+systemic vascular resistance, not including aorto-arterial segment+) R5:= SvaResistance; R5:= PenpherBisstanceVasconstriction/Vasodilator); (+systemic vascular resistance, not including aorto-arterial segment+) R5:= SvaResistance; R5:= Ponk PV: AvState(P7, R0, oldPVState); (-funcingValve+) R7:= AvState(P7, R0, oldPVState); (-funcingValve+) R7:= AvState(P7, R0, oldPVState); F10 (oldPVState=R7: (-didWVState); F10 (oldWVState=R7: (-didWVState); F10 (oldWVState=R7: (-didWVState); F10 (oldWVState=R7: (-funcingValve+); EVD(-W-signals.interpolateSignal(P4).integrator.time=0.01); LVEDP-Signals.interpolateSignal(P4).integrator.time=0.01; LVEDV-Signals.interpolateSignal(P4).integrator.time=0.01; EVD(-WN has just closed=) oldWVState=R7: (-(-mitral Valve+);</pre>
gnals.DeclStateSignal(PAP, F8, 0, 0);	$ \begin{array}{l} 112:=(P1-P2-d112d1+l12)/R1; (*12)=symmetric5(gmoid(112/112Max)+i12Max;)) (*aartic valve flow*) (*Resistance - Inductance, serial; *) (*limit shc 112B:=(P1+P2-d112d1+l2)/R1; (*12B:symmetric5(gmoid(112B/112Max)+i12ZMax;)) (*12B:sin(P1+P2), (*12B)) (*12B) $
	L20 := (Up_7=5) (+= 0L20#L23#L23#) //K4; 134 := ((0p_7+5) (-= 0344L34=) //K3; 1345 := ((0p_7+5) (-= 0344L34=) //K3; (= (Up_7+5) (+= 0344L34=) //K3; (= (Up_7+5) (+= (perphetral restance) (= (low in organs))
	(zd) = ψ = 0 / t / t / t / t / t / t / t / t / t /

- ios := (ro-rs)/Ko; (*flow across pulmonary vacular bea*) 1910 := (r9-P10)/R9; (*flow from pulmonary vein to left atrium*) 1101 := (P10-P1)/R10; (*flow across mitral valve*)



Implantation technique



Mechanical support: pressure curves & coronary flow



Summary

Computational simulation predicts

- Percutaneous LV assist (Impella 2.5) is more effective than IABP for increasing blood pressure and cardiac output, for improving coronary flow, for reduction of cardiac work, and for preload reduction
- Combining percutaneous LV assist with an IABP leads to an added benefit compared to either device alone.
- Of particular hemodynamic promise is the combination of mechanical support modalities with vasodilators
- Percutaneous LV assist is effective in complications which are not helped by IABP support, namely severe arrythmia and profound LV pump failure

Where we apply Oberon



Unsupervised image classification of echo images by automatic "elastic warping".

Schlomo V. Aschkenasy & Patrick Hunziker

Automatic image classification: An example

Sample Reference

automatic fitted sample sample & difference





e differe



How can we classify medical images ?

Optical flow equation

 $\delta I/\delta x + \delta I/\delta y + \delta I/\delta t = 0$

with I: Intensity field I = f(x,y,t)
(i.e., "objects do not disappear when they move")



Find a warping map (3D model) that minimizes the difference between a sample and a given template within smoothness constraints.

Image to Classify Template Image Deformation map Image Remaining Error Image to Classify Image Image Image Image Image Image Image to Classify Image Image

The Results (shape classification)

The Approach: warping/registration



Results: Classification by view

	Linear discriminant analysis							
Predicted Image Type								
	Class	Apex	LAX	SAX	Total			
Original	Арех	45	2	2	49			
	LAX	0	<i>24</i>	0	24			
	SAX	2	0	<i>15</i>	17			
Cross- validated	Apex	44	2	3	49			
	LAX	0	24	0	24			
	SAX	4	0	13	17			

Original: 93% (Chi2=123.8, p < .0001) 'leave-one-out': 90% (Chi2=131.1, p < .0001)

Conclusion

- A multiscale elastic registration algorithm implemented in Oberon allows the separation of shape and motion in moving datasets.
- Separating shape and motion is the basis for an automatic classification of echocardiographic images and autonomous computer vision diagnosis.
- Fast, compact code libraries make these algorithms suitable for clinical use in very large datasets.

Summary

- Oberon has proven useful for a broad spectrum of medical applications in a clinical context.
- It has been successfully applied in the fields of
 - Digital signal processing
 - Linear and multilinear algebra application
 - Image processing/analysis
 - Medical Simulation
 - Productive Telemedicine System
 - High Performance Computing
- Of particular usefulness were its following features
 - Language (simplicity, efficacy, safety, readability, maintainability)
 - Matrix language extension
 - Bluebottle server: stability, no virus threat