

Tools for Online Technical Collaboration

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Western



Menu

Appetizers

Collaboration Technical Content vs Pictures

Mains

Digital Ink Mathematical Handwriting



Previous Software Present Generation

Collaboration



Collaborative Software



Lots of Collaborative Software



Common Features

- Slide shows
- Whiteboarding
- Voice chat
- Video chat
- Image capture



Expected Enterprise Collaboration Features Circa 2013

Content	1	1		Text I Rich Media
Document & Files	g	e le	ial	MS Office PDF XML etc.
Conversation	clo	Q.	Soc	Social Depth News Feeds IM UC
Apps				Integrated Business Apps I App Stores

From http://zdnet.com/blog/hinchcliffe





"I think you should be more explicit here in step two."

from What's so Funny about Science? by Sidney Harris (1977)

Missing:

- Mathematics
- Diagrams, graphs
- Geometric figures
- Technical knowledge base
- Document markup
- Scientific software connections (Maple, Mathematica, GeoGebra, R,...)

Isn't a shared whiteboard, with the ability to save images enough????



The Treachery of Images

(La trahison des images)



 $D = \frac{1}{c} \frac{1}{dt} \frac{dl}{dt} = \frac{1}{c} \frac{1}{P} \frac{dP}{dt}$ $D^{2} = \frac{1}{P^{2}} \frac{P_{0} - P}{P} \sim \frac{1}{P^{2}} (1a)$ $D^{2} = \frac{K_{0}}{3} \frac{P_{0} - P}{P} \sim \frac{1}{K_{0}} (2a)$ D2~ 10-53 € ~ 10⁻²⁶ Pr~10⁸ L.J. t~ 10¹⁰ (10th) J

Einstein's Blackboard

- Einstein to receive honorary doctorate at Oxford, May 1931.
- Lecture at Rhodes House.
- Board retrieved and preserved by Edmund ("Ted") Bowen.
- Nice to look at, but content is **trapped**.

Digital Ink

- Location, time information, sometimes also pressure and angles.
- Capture online pen strokes, not images.
- Suitable for
 - **Recognition** algorithms
 - Semantic grouping
 - Annotation
 - Manipulation: search, transformation, archival.
- Problem: Multiple vendor-specific formats.

Ink Markup Language (InkML)

W3C Recommendation 20 September 2011

This version:

http://www.w3.org/TR/2011/REC-InkML-20110920/

Latest version:

http://www.w3.org/TR/InkML

Previous version:

http://www.w3.org/TR/2011/PR-InkML-20110510/

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Pen-Based Math

Lewegten Massenpunkstes) me J______

- Input for CAS and document processing.
- 2D editing.
- Computer-assisted collaboration.

Pen-Based Math

- Different than natural language recognition:
 - 2-D layout is a combination of writing and drawing.
 - Many similar few-stroke characters.
 - Many alphabets, used idiosyncratically.
 - Many symbols, each person uses a subset.
 - No fixed dictionary for disambiguation.

DLQDXXX $2 \dot{z}^2 \dot{z} + z = sin \omega t$

Character Recognition

- A story about a UI proposal
- A story about three statisticians
- Concentrate on character recognition
- Several projects ignore this problem

Usual Character Reco. Methods

- Smooth and re-sample data THEN
- Match against N models by sequence alignment OR
- Identify "features", such as
 - Coordinate values of sample points, number of loops, cusps, writing direction at selected points, *etc*

Use a classification method, such as

• Nearest neighbour, Subspace projection, Cluster analysis, Support Vector Machine

THEN

• Rank choices by consulting dictionary

Difficulties

- Having many similar characters (e.g. for math) means comparison against all possible symbol models is slow.
- Determining features from points
 - Requires many *ad hoc* parameters.
 - Replaces measured points with interpolations
 - It is not clear how many points to keep, and most methods depend on number of points
 - Device dependent
- What to do since there is no dictionary?
- New ideas are needed!

What the Computer Sees



What the Computer Sees



Orthogonal Series Representation

• Main idea:

Represent traces as curves, not discrete points and coordinate curves as truncated orthogonal series.

Orthogonal Series

• Start with inner product on a space of functions, e.g.

$$\langle f,g\rangle = \int_{a}^{b} f(t)g(t)w(t)dt$$

• Functions $\phi_i(t)$ give an orthogonal basis if we can write

$$f(t) = \sum_{i=0}^{\infty} f_i \phi_i(t) \quad \text{and} \quad \langle \phi_i, \phi_j \rangle = 0 \text{ if } i \neq j$$

Then $f_i = \langle f, \phi_i \rangle / \langle \phi_i, \phi_i \rangle.$

- If sum is truncated, *f* is approximated.
- Obtain orthogonal basis from any basis set, e.g. {1, t, t², ... }, by Gram-Schmidt process.

Orthogonal Series Representation

• Main idea:

Represent traces as curves, not discrete points and coordinate curves as truncated orthogonal series.

• Advantages:

- *Compact* few coefficients needed
- Geometric
 - the truncation order is a property of the character set
 - gives a natural metric on the space of characters
- Algebraic
 - properties of curves can be computed algebraically (instead of numerically using heuristic parameters)
- Device independent
 - resolution of the device is not important

Distance Between Curves

• Elastic matching:

- Approximate the variation between curves by some fn of distances between sample points.
- May be coordinate curves or curves in a jet space.
- Sequence alignment
- Interpolation ("resampling")



- Why not just calculate the area?
- This is very fast in ortho. series representation.

Distance Between Curves

$$\bar{x}(t) = x(t) + \xi(t) \qquad \xi(t) = \sum_{i=0}^{\infty} \xi_i \phi_i(t), \qquad \phi_i \text{ ortho on } [a, b] \text{ with } w(t) = 1$$

$$\bar{y}(t) = y(t) + \eta(t) \qquad \eta(t) = \sum_{i=0}^{\infty} \eta_i \phi_i(t)$$

$$\rho^2(C, \bar{C}) = \int_a^b \left[\left(x(t) - \bar{x}(t) \right)^2 + \left(y(t) - \bar{y}(t) \right)^2 \right] dt = \int_a^b [\xi(t)^2 + \eta(t)^2] dt$$

$$\approx \int_a^b \left[\sum_{i=0}^d \xi_i^2 \phi_i^2(t) + \text{cross terms} + \sum_{i=0}^d \eta_i^2 \phi_i^2(t) + \text{cross terms} \right] dt$$

$$= \sum_{i=0}^d \xi_i^2 + \sum_{i=0}^d \eta_i^2$$

- Just as accurate as elastic matching. Much less expensive.
- Linear in *d*, the degree of the approximation. < 3 *d* machine instructions (30ns) *vs* several thousand!

Problems

- Want fast response how to work while trace is being captured.
- Low RMS does not mean similar shape.



Problem 1. On-Line Coefficients

- The main problem: In handwriting recognition, the human and the computer take turns thinking and sitting idle.
- We ask: Can we do useful work while the user is writing and thereby get the answer faster after the user stops writing?
- The answer is "Yes"!

Problem 1. On-Line Coefficients

- Use modified Legendre polynomials P_i as basis on the interval [0, 1], with weight function 1.
- Collect numerical values for $f(\lambda)$ on [0, L]. $\lambda = \text{arc length}$.
 - L is not known until the pen is lifted.
- As the sample points are collected, numerically integrate the moments $\int \lambda^i f(\lambda) d\lambda$.
- After last point, compute series coefficients for f with domain and range scaled to [0,1].
 This uses a single linear transformation of the moments.

Problem 1. On-Line Coefficients

- Approach works for any inner product with linear weighting.
- This is the Hausdorff moment problem (1921), shown to be unstable by Talenti (1987).
- It is just fine, however, for the dimensions we need.

Problem 2. Shape vs Variation

- The corners are not in the right places.
- Work in jet space to force coords & derivs to be close.
- Legendre-Sobolev inner product.

$$\langle f,g \rangle_{LS} = \int_{a}^{b} f(t)g(t)dt + \mu_{1} \int_{a}^{b} f'(t)g'(t)dt + \mu_{2} \int_{a}^{b} f''(t)g''(t)dt + \cdots$$

- 1st jet space sufficient.
 - Choose μ_1 experimentally to maximize reco rate.
 - Can be also done on-line. [Golubitsky + SMW 2008, 2009]

Legendre-Sobolev Basis


Life in an Inner Product Space

- With the Legendre-Sobolev inner product we have
 - Low dimensional rep for curves (10 + 10 + 1)
 - Compact rep of samples ~ 160 bits [G+W 2009]
 - >99% linear separability => convexity of classes
 - A useful notion of distance between curves that is very fast to compute

Linear Separability



Linear Separability



Comparison of Sample to Models

- Use Euclidean distance in the coefficient space.
- Can trace through SVM-induced cells incrementally.
- Normed space for characters gives other advantages.

The Joy of Convexity

Convexity ⇒ Linear homotopies stay within a class

Add

C = (1-t)A + tB

- Can compute distance of a sample to this line
- Distance to **convex hull** of nearest neighbors in class gives best recognition [Golubitsky+SMW 2009,2010]





Choosing between Alternatives



Training

• Using CHKNN allows training with relatively few samples. (Dozens vs Thousands per class)

Error Rates as Fn of Distance



Combining with Statistical Info

- Empirical confidence on classifiers allows geometric recognition of isolated symbols to be combined with statistical methods.
- Domain-specific *n*-gram information:
 - Research mathematics 20,000 articles from arXiv [MKM -- So+SMW 2005]
 - 2nd year engineering math most popular textbooks [DAS -- SMW 2008]
 - Inverse problem identifying area via *n*-gram freq! [DML -- SMW 2008]



Baseline Estimation

• Figure out baseline from the characters, rather than the other way around, which is more ususal.

• We can locate some important features by identifying special points.



We refer to a point such as this, that determines the height of a metric line, as a *determining point*.

Baseline Estimation

- Juxtaposition ambiguity Pg Pg Pg Pg Pg Pg P9 Pq pq pg p9
- Handwriting neatening

$$\alpha_1 \chi^2 + \alpha_2 \rightarrow \alpha_1 \chi^2 + \alpha_2$$

Average Symbol

• The average symbol of a set of known samples for a class can be computed as the average point in the functional space, n

$$\bar{C} = \sum_{i=1}^{n} C_i / n$$



Deriving from a Reference Symbol



Using Homotopy

• Some samples are far away from the reference symbol.



Average



• We use a homotopy between the reference symbol and the target sample in a multi-step method.



• 2000 Cross Pad:





• 2002 Pocket PC:





• 2002-2008 Tablet PC:





• 2008-2013 Java Application:





InkChat (Java Version)

• Skype and GTalk add-on to the Java application.



Problems

- Requires installation:
 - Big hassle for someone to use only once in a while or on all their machines.
- Limited portability:
 - Users expect versions on Android, iOS, Windows, Mac OSX, Linux, etc...
 - Incompatible software bases
 - Flakey, moving APIs
- Need to support multiple devices.
 - Nowadays a single user will want to work across many devices.

Solution

•Use browser infrastructure.

Solution

- •Use browser infrastructure.
- JavaScript is not a great language for large projects, but....
 - It is ubiquitous: Telephones, tablets, laptops, ...
 - Libraries for many UI elements
 - Our new recognition algorithms are fast enough ③
- Rapid development:
 - Prototype developed in 3 months by 3 students.

Current Generation



Desktop

Telephone

Current Generation



Simple Interface with device-adapted menus



Simple Interface with device-adapted menus



Ink Controls





Collaboration: Multiple Users Connected to Same URI



Collaboration: Different Viewports from Different Devices



Collaboration: Pointers for Discussions



Collaboration: Document Annotation



Collaboration: Google Hangout Embedding





Cloud Integration

- Save or load files to cloud storage
 - DropBox
 - Google Keep
 - Others possible
- Previous work to store user profiles
 - Save cloud of ground-truth labelled symbols (corrected/accepted)
- Future work to store user-defined brushes










1 - Notification of Participant event (object creation / deletion / movement history navigation / page change / etc)

2 - Notification of Server Event (Passed on from another client or result of server configuration)

Architectural Direction $\left(H \right)$ a+b=c / (5b) 1 (5a) 1 - New stroke / event 2 - Distribute to recognizers for screen area Participant Logic 3 - Connect object space 4 - Return ranked results Participant Object 5 - a Update presentation Space b Update server (4)(2)2(4)4` Rec2 RecN Rec1 \mathbb{N}^{3} R Z 3 1 3

Application Web Site



Conclusions

- Technical collaboration requires tools not found in the business setting.
- Drawing, mathematics and scientific documents are in the work flow.
- The treachery of images.
- Needed:
 - Math handwriting recognition.
 - Easy geometry and diagrams.
- Even a little goes a long way....

- Document mark up.
- APIs to scientific software.
- ... there is a lot of opportunity for future development.



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