

Document Image Analysis

State of the Art
and
Technology Roadmap

March, 2008

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Every Document Tells a Story...

Enterprise
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ATNO, LAFAYETTE, CT.
805-553-7426

LA HOJA Martes, 27 de enero del 2004

BUSCAN UN ACUERDO DE COOPERACIÓN Ecuador y Japón intercambian experiencias sobre volcanes

El objetivo de los encuentros técnicos es que se firme un acuerdo de cooperación que incluya la donación de una nueva tecnología de monitoreo para los volcanes Cotopaxi y Tungurahua.

Al fin de mejorar la vigilancia en los volcanes Tungurahua y Cotopaxi, técnicos del Instituto Geofísico de la Politécnica Nacional (IGEPN) y una misión de expertos del Japón comparten experiencias sobre las nuevas tecnologías.

Según el director del IGEPN, Hugo Yépez, la visita de los japoneses responde a un pedido del Ecuador para lograr un convenio de cooperación que podría incluir la transferencia de equipos para la vigilancia de volcanes.

"Una misión de japoneses evaluó la propuesta hecha por el país y otra misión se encuentra evaluando las condiciones de campo para la vigilancia de los volcanes", confirmó Yépez, quien espera que luego del trabajo técnico se logre la firma definitiva del convenio.

De lograrse un acuerdo, el Gobierno de Japón donaría sismógrafos de banda ancha y micrófonos con sensores acústicos

Los Técnicos del Instituto

que responden a muy bajas frecuencias.

Estos equipos, dentro de lo previsto, serían instalados en los volcanes Cotopaxi y Tungurahua, que mantienen distintos tipos de actividad y que, actualmente son vigilados por una red de sismógrafos de baja frecuencia.

VIGILANCIA INTEGRAL

Tras dejar claro que todo depende de la aprobación del Gobierno japonés, Yépez explicó las ventajas que tendría Ecuador con la instalación de los nuevos equipos, los cuales se usan para la vigilancia de

ría el uso de aparatos a una cñaes producidas

Los equipos ven una banda permitirá vigilar amplio de señal

Esta instalación la incorporación en el registro y que emiten los v no como exte

Sudan Embassy In South Africa
ESZA/OGX/H2CF
01/09/2006
Our Ref:..... Your Ref:..... Date:.....
Dear Sir,
FROM THE DESK OF HONOURABLE MINISTER OF ENERGY AND MINERAL RESOURCES REPUBLIC OF SUDAN: DR AWAD AL-JAZ
SOURCE FOR AGENT/PURCHASE OF MACHINERY
I enclose my communication through email, on an issue as important as this one. We shall communicate as soon as I receive a response from you. I write to inform you on behalf of the Director of Contracts and Finance Allocations country (Republic of Sudan).
The Machinery are available and can be obtained in... I have therefore been directed to inquire if you... that this mail is purely for Republic of...
The only source of income for the... are taking advantage of the situation to... even execute contracts for individuals,
I am sending you by forwarding to me, your detail... numbers for easy communication. The... have agreed to highly compensate you in your participation this transaction; nonetheless,
I am using my above stated email address or... for any communication for seeing this transaction through. Your... a positive response from you, I shall give you a complete
I am enclosing...
Dr. Awad al-Jaz
Energy and Mining ministry
Government House, Khartoum
Republic of Sudan
Be notified that The Minister have a personal interest in this transaction; as a result I want my own company to handle this transaction. Due to the current war in the Darfur region, our head office has been moved to Johannesburg, South Africa due to security reasons and enabling environment. Be informed that all correspondence and meetings will be held in Johannesburg till conclusive stages which would be held in Khartoum at the Government Lodge.
SIGNED
South African Contact: ...
Contact: Mr. Monday Sae Kumba: Third Secretary
Contact Telephone and Fax numbers to send you upon reply due to security reasons.

Ervin Leasing Company
Dept. 77228
P.O. Box 77000
Detroit, MI 48277-0228 (734) 332-5400
INVOICE DATE 04/05/2001

CONTRACT NUMBER	PAYMENT NUMBER	DESCRIPTION	CONTRACT PAYMENT	SALES/USE TAX	LATE CHARGES	AMOUNT
						956082) \$435.72
						001 \$435.72
						251.76
						Subtotal 251.76
						143.96
						Subtotal 143.96
						void late charges. **
						228 ***
						TOTAL DUE 395.72
						BIT ON YOUR ACCOUNT MI, MI 48277-0228
						act # Amt Due 251.76
						143.96
						al Due \$ 395.72
						Date: 04/20/2001
						#2510001 0000 3

mer...

Hugo Yopez hosted some Japanese

Sudan procured oil drilling equipment...

Outline

1. The classical document recognition pipeline

- image preprocessing
- page segmentation
- text recognition
 - OCR
 - Handwriting
- page layout analysis
- text analysis

State of the art:
It all depends...

2. Why is this so hard?

3. Trends

- What is industry up to?
 - the paper/digital divide
- What are academics up to?
 - digital libraries
 - camera document imaging

4. Structure of the Field

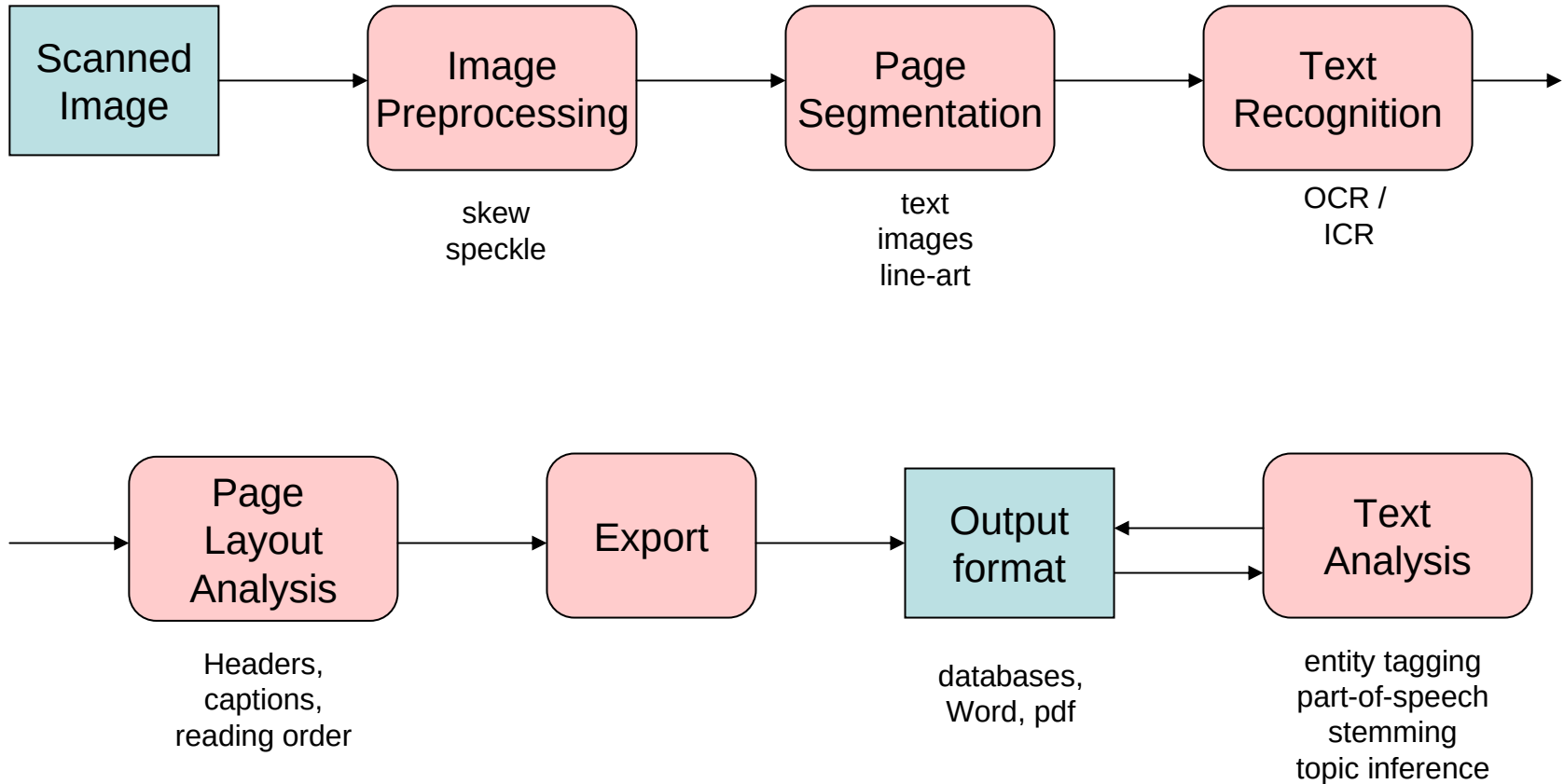
- Machine learning
- Where is research done?

5. Roadmap: pressing problems and opportunities

Acknowledgements

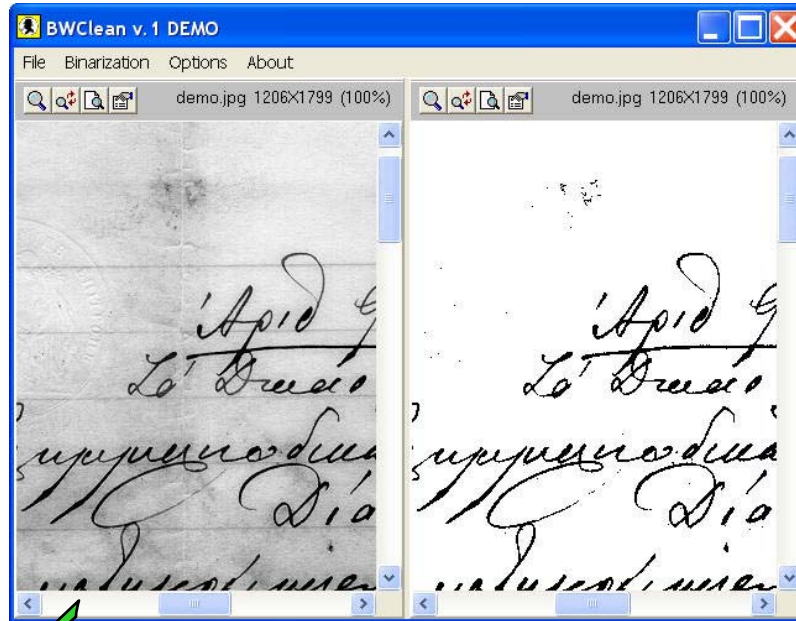
Dave Doermann, U. Maryland
R. Manmatha, U. Mass
Dan Lopresti, Lehigh U.
George Nagy, RPI
Yann LeCunn, NYU Courant Inst.
Marco Bressan, XRCE
Prateek Sarkar, PARC
Tracy King, PARC
Danny Bobrow, PARC
Stu Card, PARC

The Classical Document Recognition Pipeline

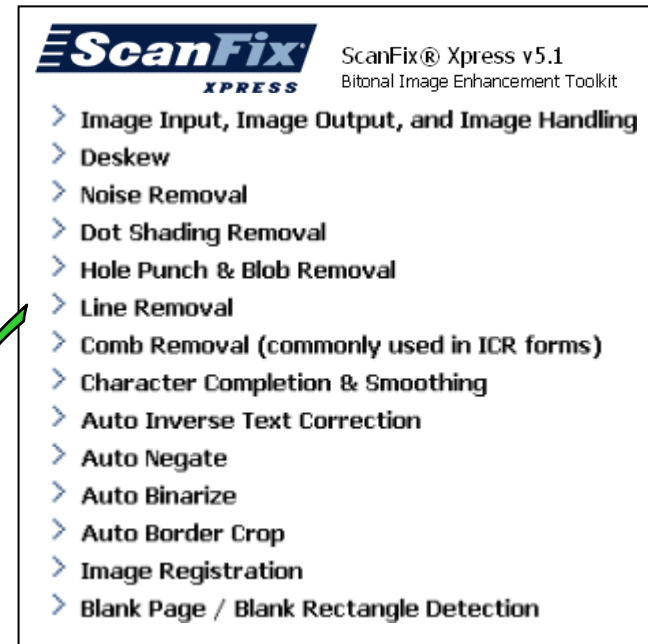


The Classical Document Recognition Pipeline

Image Preprocessing



✓ Adaptive thresholding for binarization



Dozens of commercial products

Not available: ✗

- separate multiple content layers
- non-uniform skew
- detect multiple documents in one scan image



The Classical Document Recognition Pipeline

Page Segmentation

Purpose:

- identify zones of text, image, line-art, page graphics (separators)
- used widely to improve compression and image quality of scanners/copiers

Main approaches:

- recursive decomposition by analyzing whitespace runs
- image morphology to form blobs of like image material



The Gist of Translation

How long will it be before machines make the Web multilingual?

BY JENNIFER L. SCHENKER PARIS

MIRKO PITTI, A SWISS ENGINEER IN PARIS, is helping and his could go to the Gap's website and find a description of the clothes—and an order form—in their native language. That day isn't far off.

Machine-language translation—a process that uses computer software to translate text from one language into another—is maturing precisely at the time when the need for it, particularly on international websites, is escalating. Analysts figure that by 2005 about three-quarters of the 1 billion people communicating and doing business online will live outside the U.S., and most of them won't speak English.

Late last month Autodesk, a U.S. software company, began offering machine-generated translation to its European customers at 50% less than it costs to have humans do the job. In its first application of the software, Autodesk is providing French and Spanish versions of 5,000 online customer-support documents—the ones that tell you what to do when you run into technical problems. German and Italian will be added in about a month, with Asian languages to follow, says Mirko Pitti, a process analyst at Autodesk's global center for localizing content in Neuchâtel, Switzerland.

The company supplies Autodesk's translation software to Sysran, which is based in Soire-sous.

CAN WE TALK? Autodesk's Mirko Pitti says German, Italian and Asian languages are next.

Montmorency, France. (www.sysran.com). It makes software that can translate more than 20 language pairs and provides the technology behind the free online translation services offered by AltaVista's Babelfish, Google, Lycos and AOL, as well as a wireless portal run by Oracle.

The quality of these free translation services tends to be uneven because the subject matter is so diverse and the process relatively imprecise. (It is called "gisting," because it yields translations that provide the gist of a document.) What Sysran offers Autodesk—and other multinational companies—is improved translation software customized for the content of the website. "Multinational organizations trying to provide quality service to all their customers, business partners and employees will want to have this capability," says Steve McClure, a research vice president at the technology consultancy IDC in Framingham, Massachusetts. Given Sysran's head start, IDC says it holds the No. 1 position worldwide for machine-language translation, ahead of IBM and Belgium's Mendex.

Machine translation has long been an elusive goal of computer experts. Like its human counterpart, the software has to take into account the grammatical structure of each language; it can't just give a substitution for every word unless it understands all of the words in a given sentence and how one influences the other. Early attempts at decoding natural languages through mathematical techniques using closet-sized computers produced poor results. Then Peter Yone, a linguistics researcher who began his work in 1957 at the California Institute of Technology, introduced linguistic rules into the process. He founded Sysran in 1968. Among its first clients was the U.S. Air Force, which asked the company to develop a Russian-into-English machine-translation system. Later work for NASA caught the attention of the European Commission, which asked Sysran to develop systems to translate European languages. In 1986 the company was purchased by Gecot, a French maker of industrial valves and fittings. Sysran, which trades on the Nouveau Marché, last year had profits of \$376,000 on revenue of \$3.5 million.

For decades Sysran's machine-translation technology was available only to U.S. and European government agencies and other public institutions via mainframe computers. Today its software runs in many different hardware platforms—including the type of Linux and Unix servers used by multinational corporations. At Autodesk the Sysran software will not replace the people manning the multilingual help desk in Neuchâtel, says Pitti, but it's likely to make their jobs easier.

Sysran has other multinational clients, including Ford and Price Waterhouse Coopers, that use the technology internally. But to create a mass market, Sysran will need a lot more customers like Autodesk, which Sysran CEO Dimitris Sabatakakis calls a "showcase client," putting the software on their external websites. If analysts are right, translating customer-support information is a significant first step, after which e-commerce applications should follow. Beyond that, the World Wide Web will be not just multilingual but multilingual as well.

VERSION 1.0 (REVISED)

The Classical Document Recognition Pipeline

Optical Character Recognition

Behind every male leader, there's a female calling the shots. At least that's true in one type of African fish, according to research that contradicts previous notions about male-determined hierarchy.

The cichlid fish *Neolamprologus pulcher* lives in groups in Lake Tanganyika in southeastern Africa. Each social group contains one breeding pair and about 20 other males and females that look after the young and ward off predators. When the alpha male dies or is displaced, another comes forward to breed with the alpha female.

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(ABBYY FineReader)

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RESOURCES REPUBLIC OF SUDAN: DR AWAD AL-JAZ

SOURCE FOR AGENT/PURCHASE OF MACHINERY



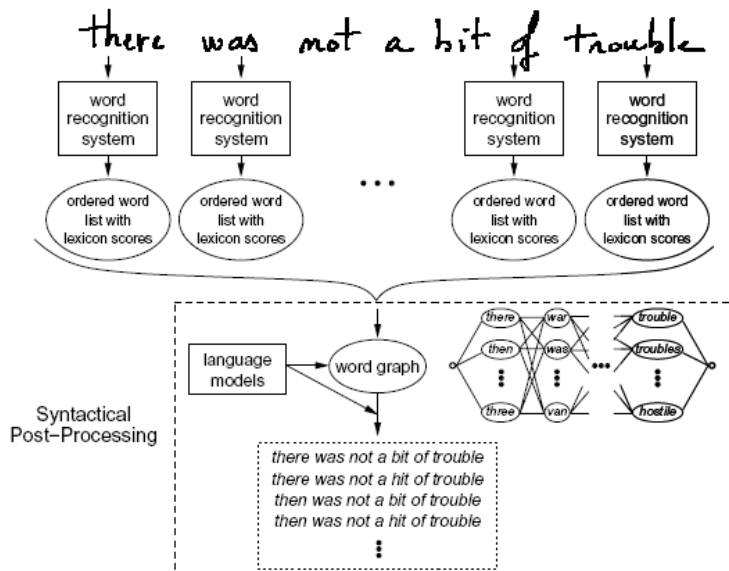
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KFSC>1IK<T.S HICniBI.K Of SI DAN: OR A WAI) AL-JAZ
SOURCE FOR AOENT/PIKCHASK OF MACHI.N KRY

State of OCR:

- commercial market mature
- results depend heavily on image quality
- “up to” 99.9% accuracy: ~ 2 errors/page
- non-Roman alphabets still lagging

The Classical Document Recognition Pipeline

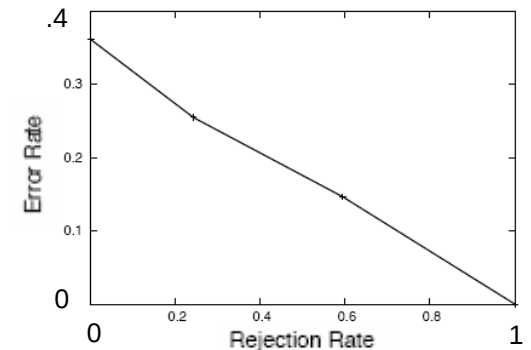
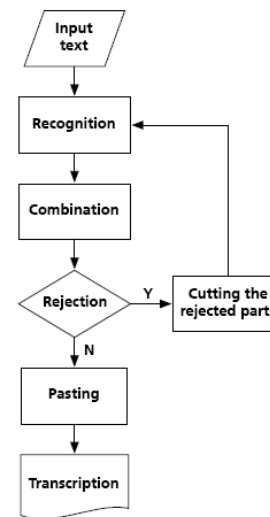
Handwriting Recognition



Brown corpus Model	Word rec. rate
Baseline	83.53 %
Bigram	91.72 %
Trigram	92.30. %

State of handwriting recognition:

- Online writer-trained HW rec. is very good
- Offline general HW rec. unusable for transcription, but may support some search and word-spotting applications
- Isolated numerals very good: > 99%
- Finding and segmenting handwriting still an issue
- Much work to be done on non-English scripts

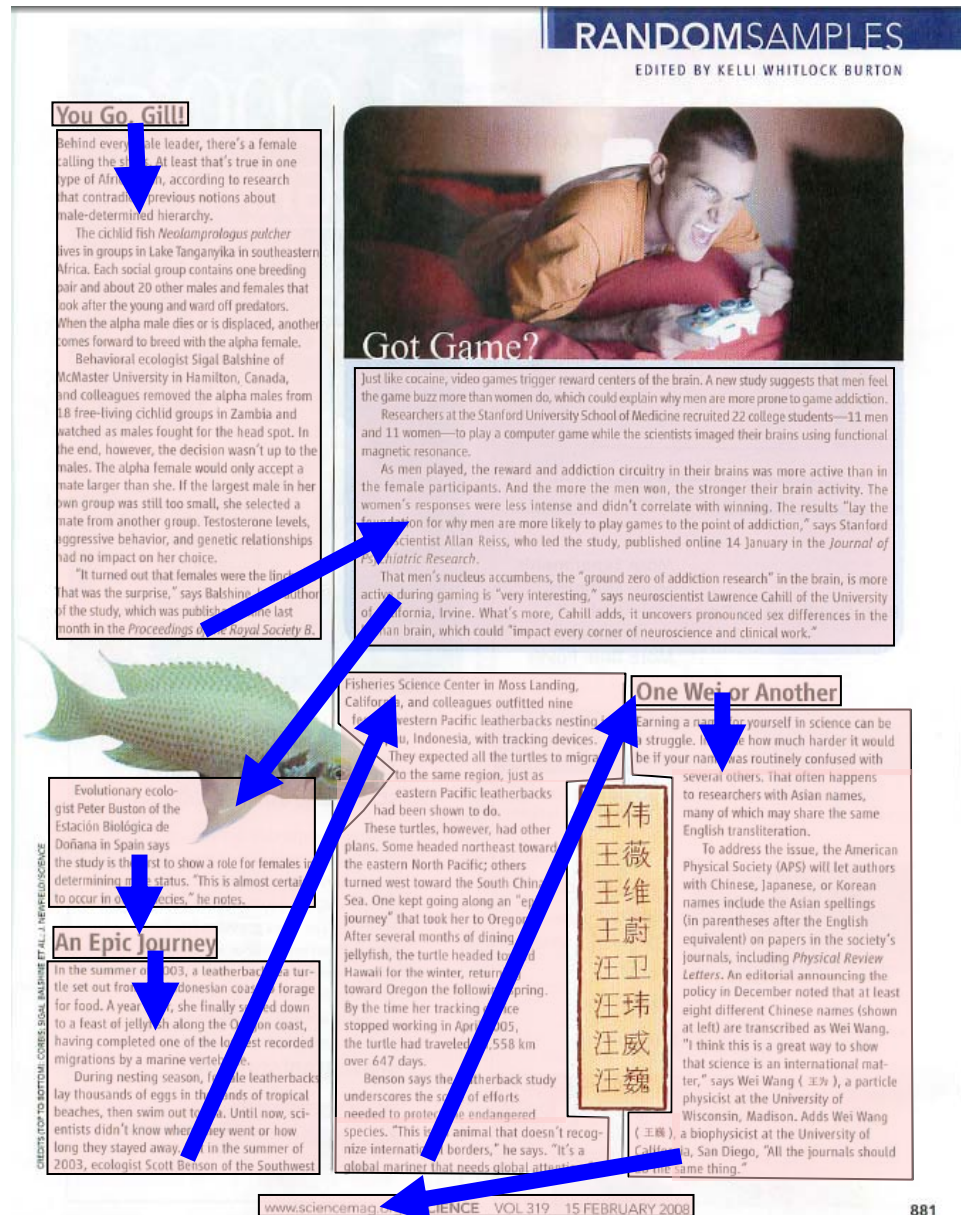


Word recognition error vs. reject

The Classical Document Recognition Pipeline

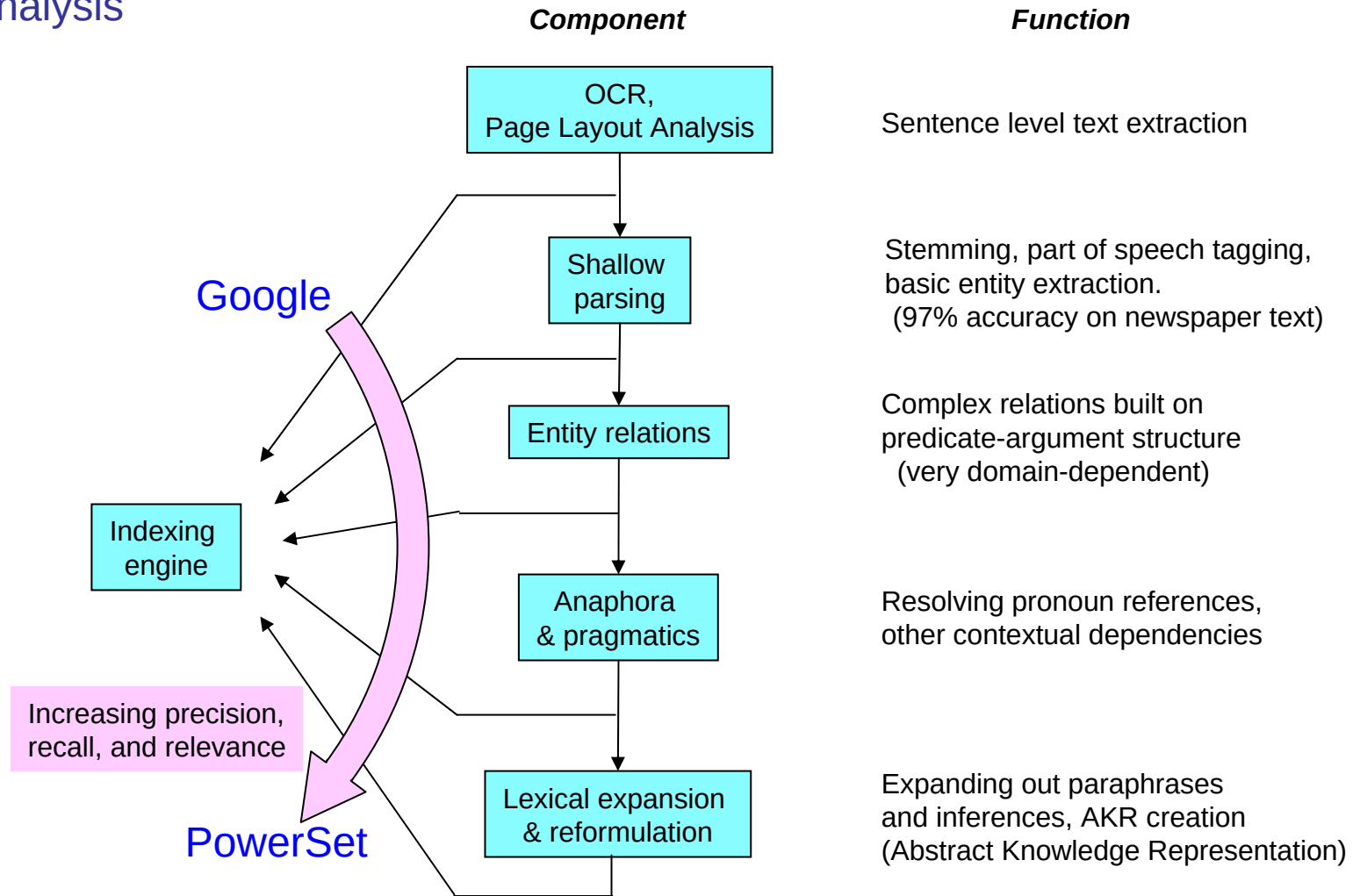
Page Layout Analysis

- Reading order
- Functional Role Labeling (Zone typing)
e.g. title, header, caption, page number.
- Table parsing
- Current methods mostly rule-based, heuristic, error-prone.
- Current formulations are in terms of
 - document logical structure vs.
 - document layout structure.
- Lack general theories and formulations for combining weak evidence.
- Lack ontologies of layout elements.
- Lack taxonomies of document genres and layout patterns.



The Classical Document Recognition Pipeline

Text Analysis



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- Machine learning
- Where is research done?

5. Roadmap: pressing problems and opportunities

What Makes Document Recognition Difficult?

- variability

- 100000s of journal, magazine, newspaper formats

- unpredictability, exceptions

- handwritten annotations, post-it notes, strange fonts, languages
- figures, graphics can include anything

- context

- meaning of glyphs, graphics, whitespace depend on---and define---genre and style

- You can engineer a solution to any narrowly constrained problem.

- Real applications seldom obey narrow problem constraints.

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Trends: Commercial Document Recognition

Driving Force: Bridge the paper/digital divide



**Is running your business
getting in the way of running
your business?**

Then get QuickBooks and spend
less time on your paperwork and
more time on your business. It's
the easiest way to:

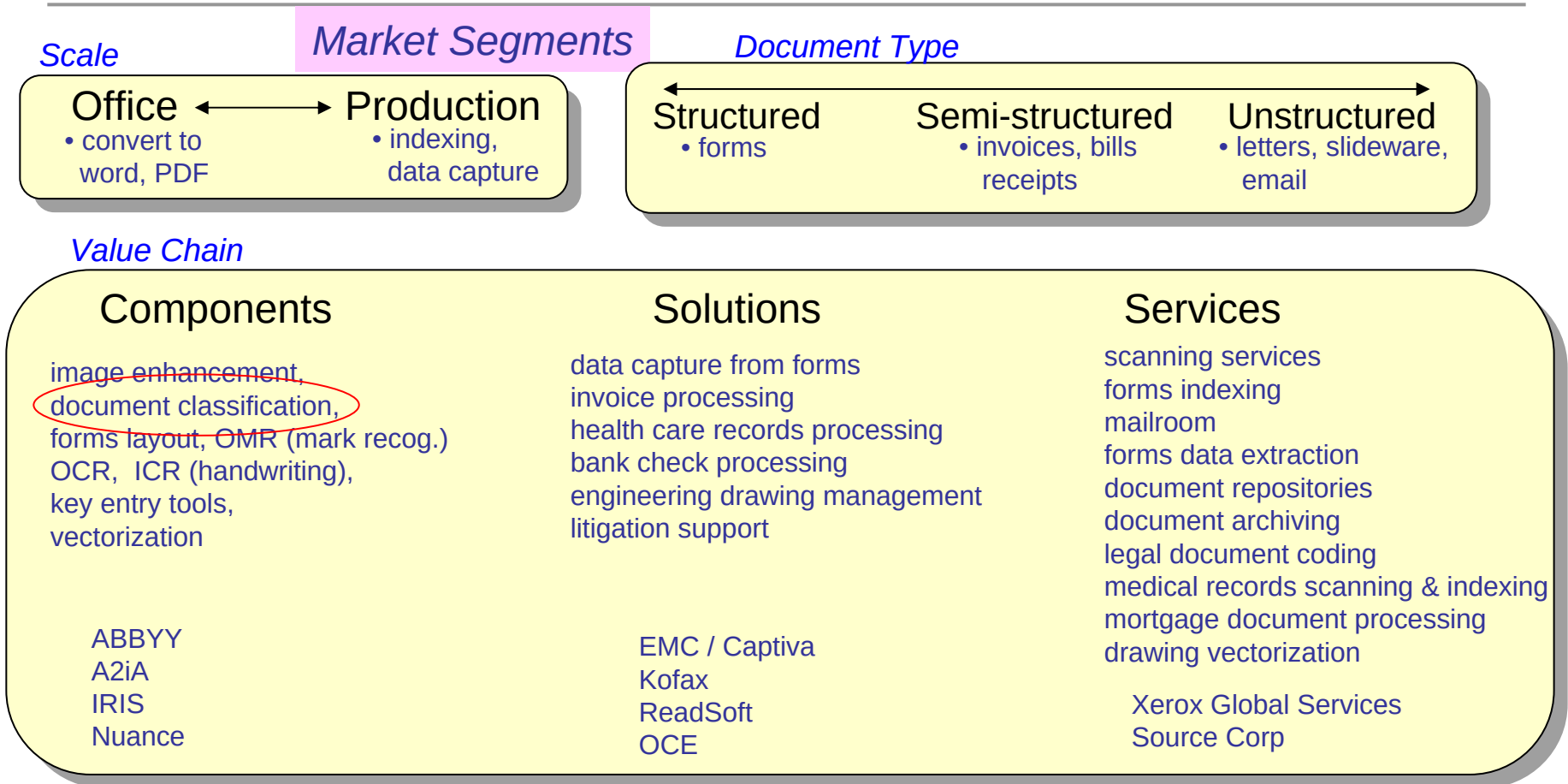
CREATE INVOICES

PAY BILLS

TRACK SALES

www.QuickBooks.com

Trends: Commercial Document Recognition



Accessible \$ ≠ connect the dots

- large volume of predictable documents
 - offshoring
- value to customer

- large volume of variable documents
 - human assist
- need flexible access for fishing

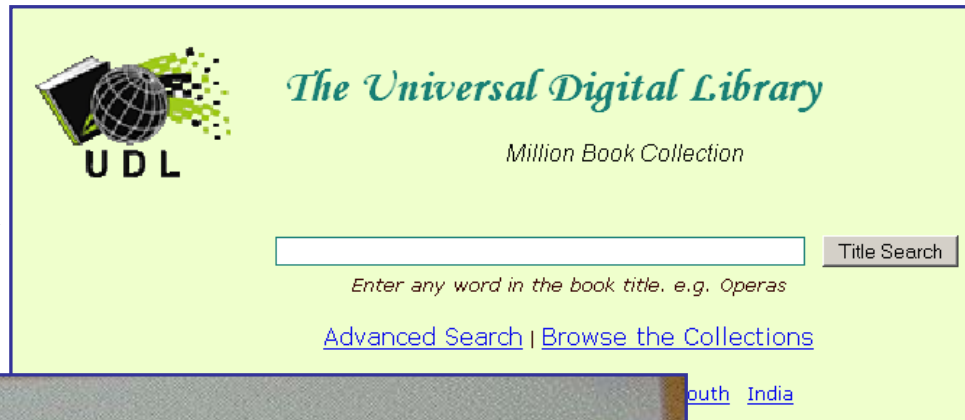
Trends: Academic Research

Drivers:

- Digital libraries

Google books, Million book project,
historical document mining, geneology

- Cameras
- Web data capture
- Pen computing



Trends: Academic Research

Drivers:

- Digital libraries
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Topics:

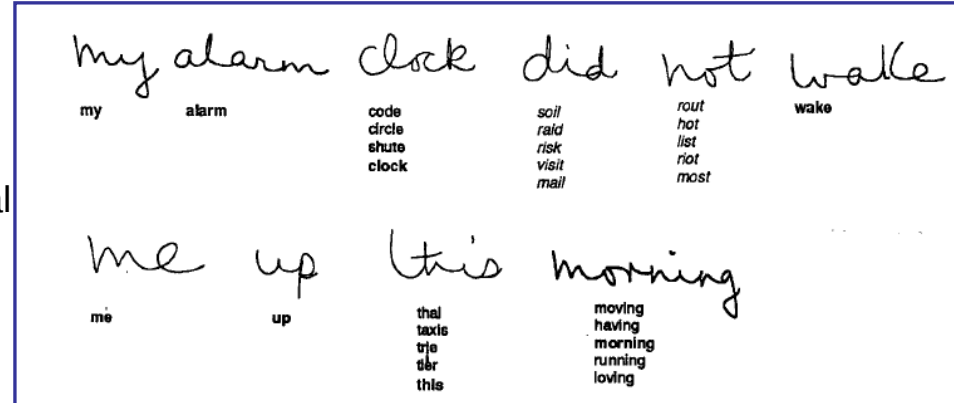
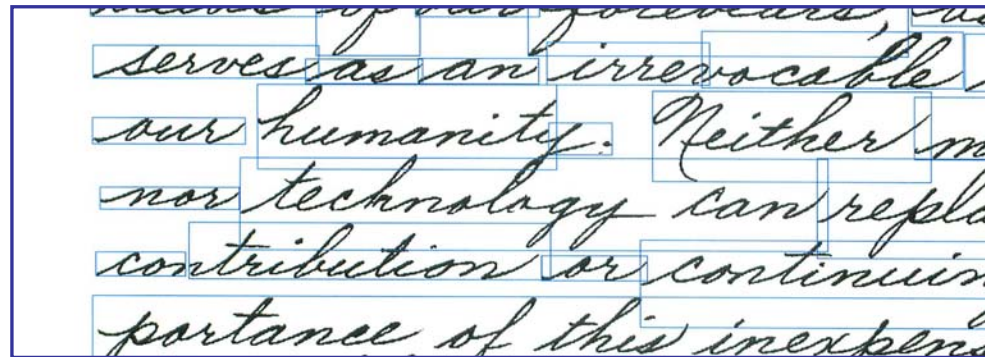
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 - parsing and segmentation
 - lexical constraints
 - learning and inference methods
- OCR on degraded documents
- languages, non-roman alphabets
- word spotting
- image preprocessing for lighting, deformation, degraded sources
- scene text - finding and reading
- writer ID / signature verification
- document image matching and retrieval
- layout modeling
- graphics recognition
- maps

Trends: Academic Research

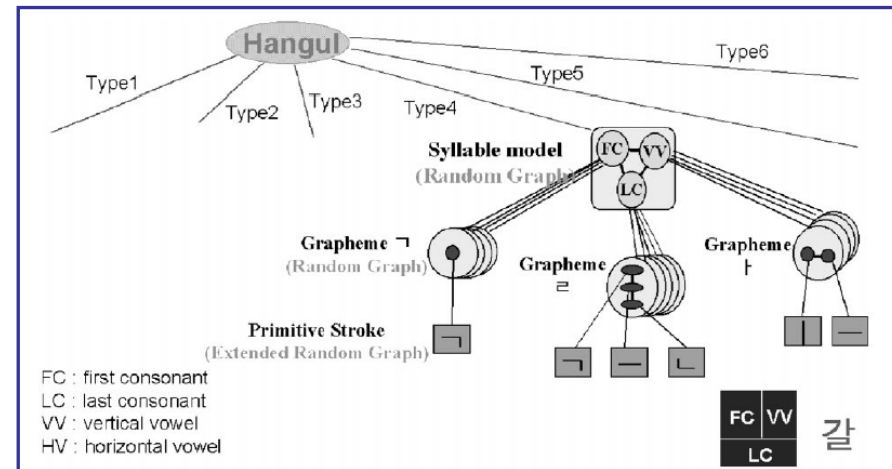
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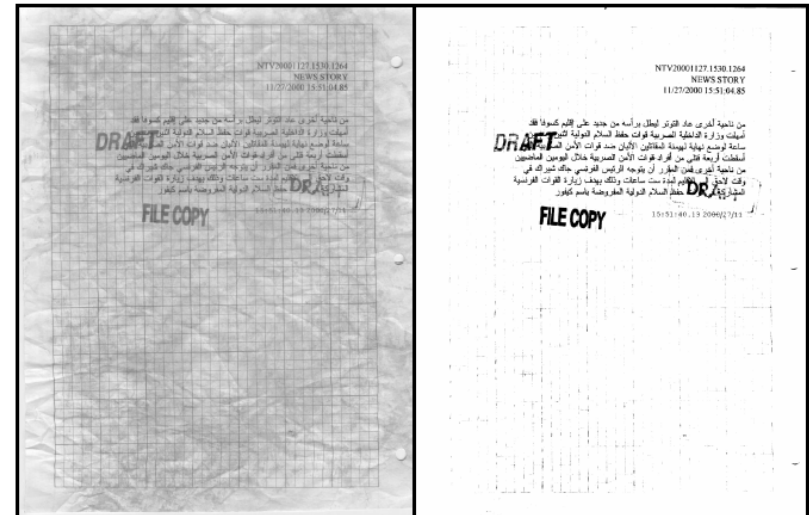
- The “face recognition” of document image analysis.



Trends: Academic Research

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Optimizing OCR accuracy for bi-tonal, noisy scans of degraded Arabic documents

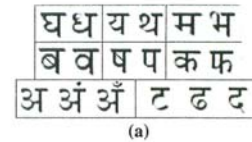
Paul Herceg, Ben Huyck, Chris Johnson, Linda Van Guilder, Amlan Kundu*
The MITRE Corporation, 7515 Colshire Drive, McLean, VA 22102-7508

- Lots of very helpful image processing can be done.
- The problem is knowing what to apply to any given image.

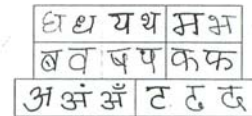
Trends: Academic Research

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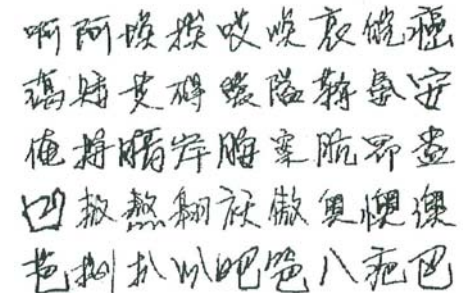
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(a)



Devnagari



Chinese



Arabic



Bangla

Base Sound	Orders						
	1 st (ā)	2 nd (u)	3 rd (i)	4 th (a)	5 th (e)	6 th (ā)	7 th (o)
1	h	ḥ	ẖ	ẖ	ẖ	ḥ	ḥ
2	l	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
3	h	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
4	m	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
5	s	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
6	r	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
7	s	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
8	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
9	q	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
10	b	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
31	p'	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
32	f	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
33	p	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ
34	v	ḥ	ḥ	ḥ	ḥ	ḥ	ḥ

Ethiopic

9	10	11	12	13	14	15	16	17
18	19	20	21	22	23	24	25	26
27	28	29	30	31	32	33	34	35
36	37	38	39	40	41	42	43	44
45	46	47	48	49	50	51	52	53
54	55	56	57	58	59	60	61	62
63	64	65	66	67	68	69	70	71
72	73	74	75	76	77	78	79	80
81	82	83						

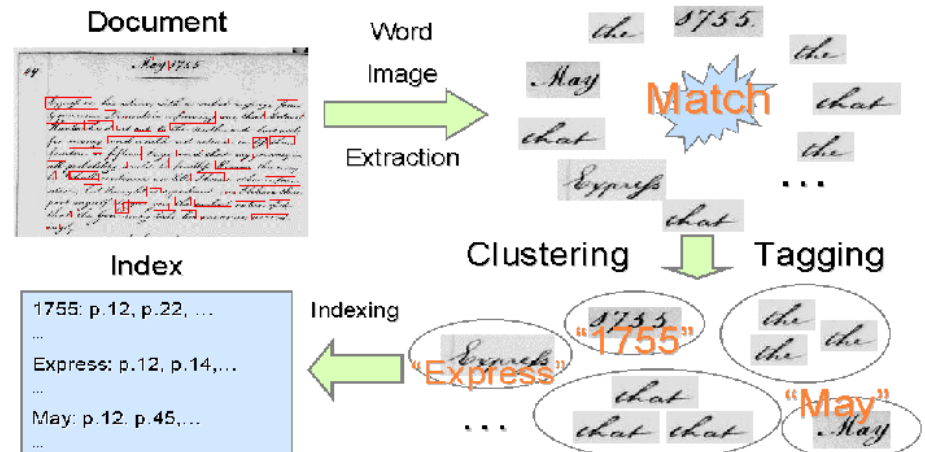
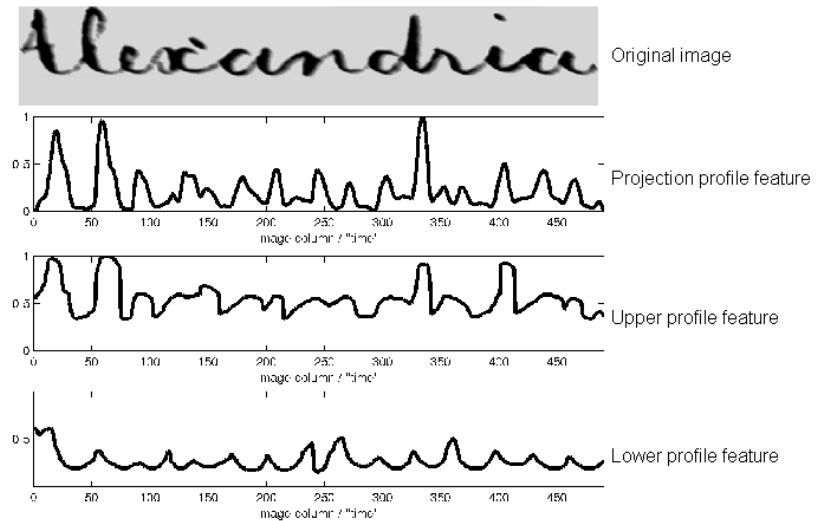
Tamil

- Lots of activity overseas is training a world full of document recognition practitioners.

Trends: Academic Research

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- Task-driven workaround for difficulty of handwriting recognition & OCR on difficult documents.

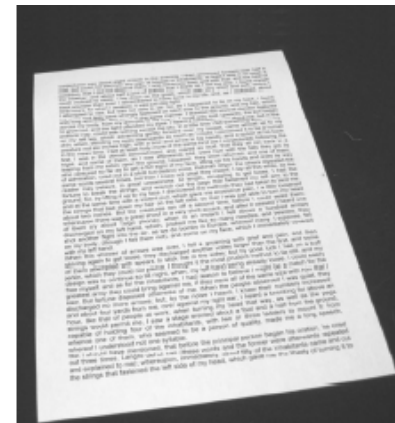
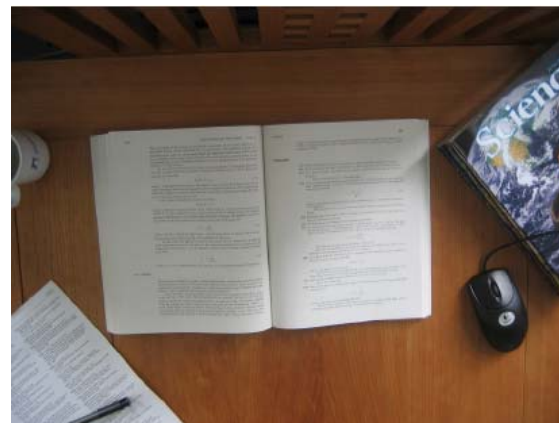
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Landon, Lin, Seales, U. Kentucky



Lampert, Braun, Ulges, Keysers, Breuel, IUPR, Germany

- Ubiquitous computing:
potential to radically
transform work.

Trends: Academic Research

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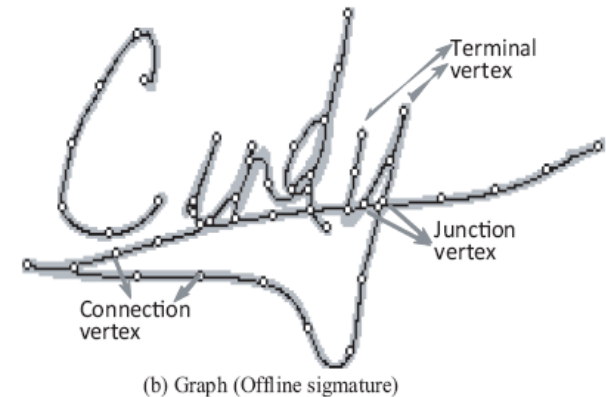
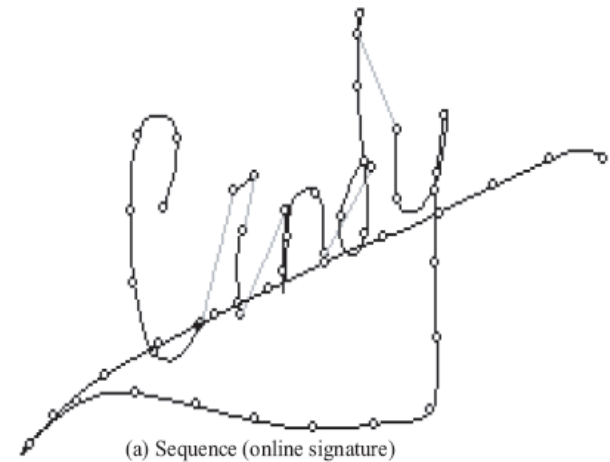
- Expect language translation in Japanese cameras soon...

Silapachote, Weinman, Hanson, Weissy, Mattar, U Mass

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Qiao, Liu, Tang, Chinese U. of Hong Kong

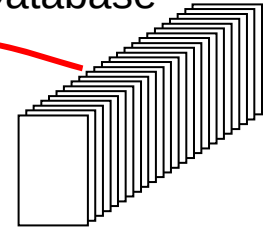
- Useful for filtering obvious cases, but even experts argue.

Trends: Academic Research

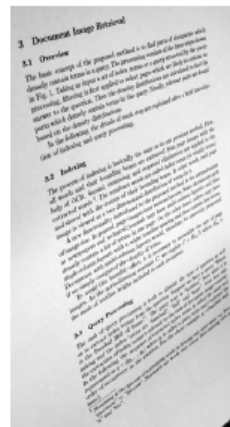
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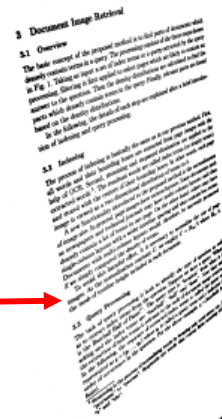
Document
Image
Database



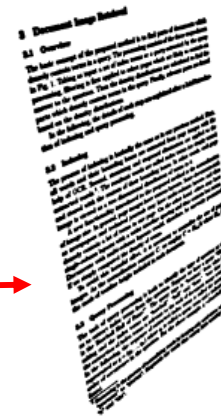
fingerprint
index



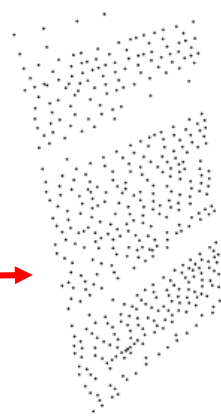
(a) Input image.



(b) Binarized image.



(c) Connected components.



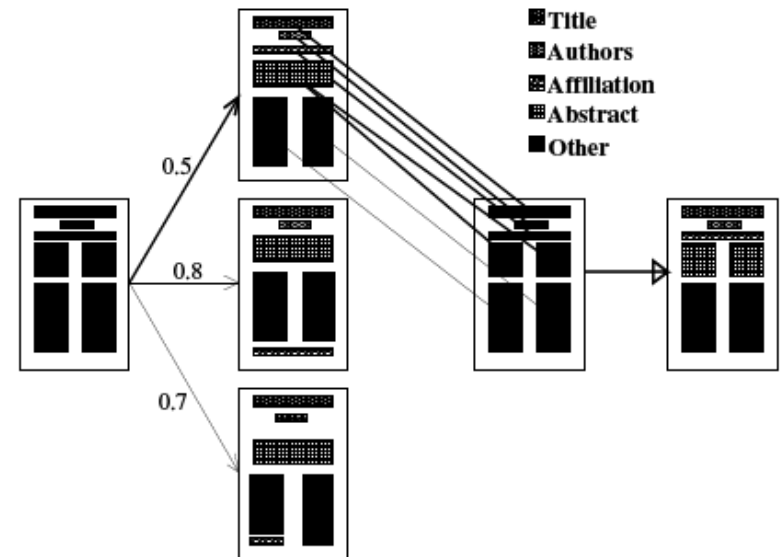
(d) Feature points.

- Breakthrough is in the lab now, looking for customers.

Trends: Academic Research

Topics:

- handwriting
 - parsing and segmentation
 - lexical constraints
 - learning and inference methods
- OCR on degraded documents
- languages, non-roman alphabets
- word spotting
- image preprocessing for lighting, deformation, degraded sources
- scene text - finding and reading
- writer ID / signature verification
- document image matching and retrieval
- • **layout modeling**
- graphics recognition
- maps



Beusekom, Keysers, Shafait, Breuel, DFKI, Germany

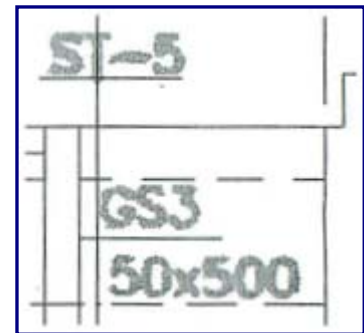
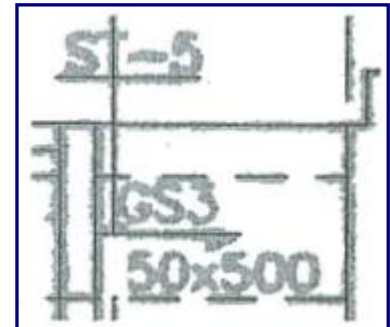
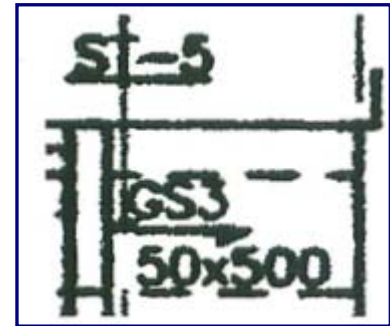
- No one has a clue here.

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- A whole field unto itself, no longer stuck at vectorization.



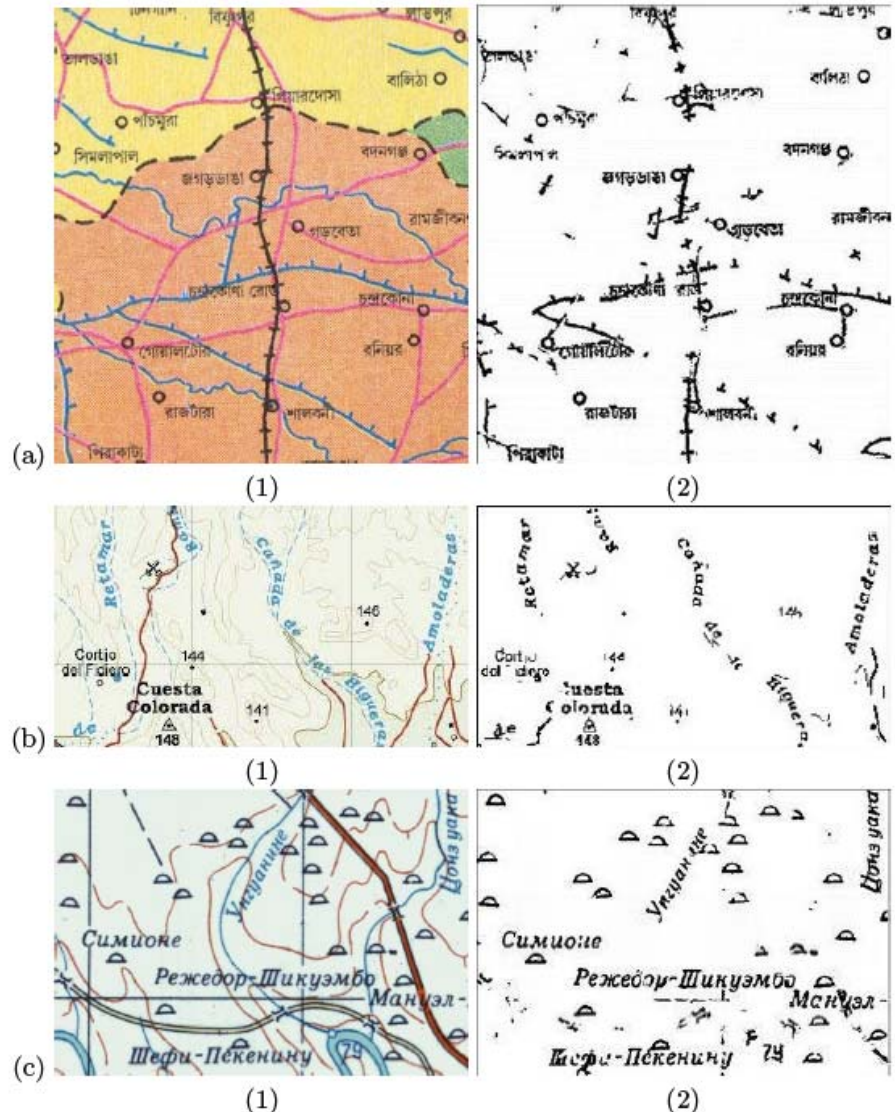
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- Incredibly challenging, human assist seems the only way forward for now.



Outline

1. The classical document recognition pipeline

- image preprocessing
- page segmentation
- text recognition
 - OCR
 - Handwriting
- page layout analysis
- text analysis

2. Why is this so hard?

3. Trends

- What is industry up to?
 - the paper/digital divide
- What are academics up to?
 - digital libraries
 - camera document imaging



4. Structure of the Field

- Machine learning
- Where is research done?

5. Roadmap: pressing problems and opportunities

Machine Learning in Document Recognition

- Primary use: adjusting parameters of fixed models based on statistics of data
 - character classification
 - document type classification
 - HMMs for string recognition (e.g. handwritten words)
- } Kernel / Support Vector Machine
Neural Net / Multilayer Perceptron
Decision Tree

Data is critical

- large quantities
 - preferably real, not simulated
 - groundtruthing
 - competitive advantage in industry
 - data obfuscation
- Exploratory:
 - learning of features as well as classes (e.g. deep belief networks)
 - unsupervised e.g. clustering to learn novel glyph sets
 - build structural models of layout patterns
 - MRF, CRF models for page segmentation
 - discriminative learning for structured output

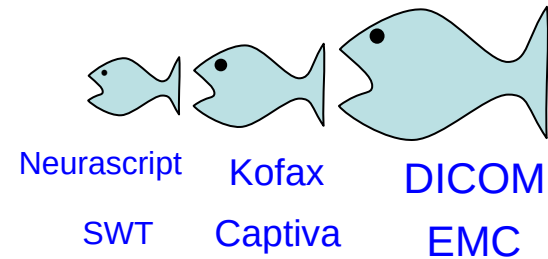
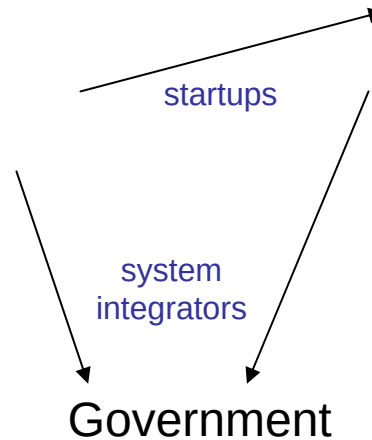
Where is Progress in Document Recognition?

Universities

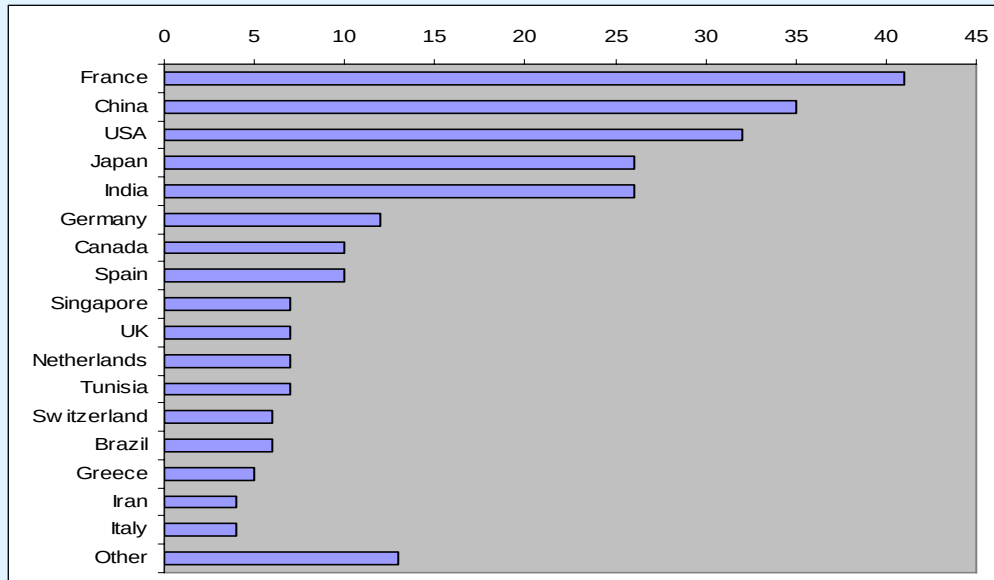
- harvest advances in computer vision, machine learning, natural language, digital libraries
- imaginative but need connection to real-world problems

Industry

- short-term outlook
- secretive
- R&D by acquisition



ICDAR 2007 papers by country



Microsoft Research

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5. Roadmap:

- bottom line state of the art
- pressing problems and opportunities

Bottom Line State of the Art

Available Commercially

Current research / Could be engineered

Future Research



Office OCR

- convert clean office memo for pdf, editing

- extract main fields and topics of memos in company correspondence

- scan and sensibly organize the contents of an office (mixture of memos, published material, receipts, notes, invoices, etc.)

Books

- scan and OCR books for search

- extract structured representations for Russian language proceedings of one publication

- parse all encyclopedias since 1900

Handwriting

- index online handwritten notes from a Tablet PC

- index and search names and places in neatly handwritten papers

- transcribe the notes off a typical conference room whiteboard

Cameras and scene text

- spot and translate street signs in camera images

- translate a posted bus schedule

Database search

- find potential document duplicates through text string matching (depends on good OCR)

- find document matches in a database from a low-res camera image

- automatically construct a version history of documents scanned multiple times over their authoring cycle

Roadmap: Pressing Problems and Opportunities

Topics

- structural models for document layout
 - combined segmentation and labeling
 - representations for variation, patterns, degrees of freedom
 - learning
 - inference
- principles for document genre
 - multidimensional types and taxonomies
 - multivalent definitions of similarity
 - learn from corpora
- document collections
 - relations among documents and items within a “case”
 - probabilistic logic – combine rules with statistics

Process

- build data sets reflecting problems of interests
- build community
 - workshops
 - competitions
 - evaluation methodologies
 - “a few good researchers”
- leverage cross-discipline interaction
 - computer vision, machine learning, natural language, Artificial Intelligence
- mine open source

saund@parc.com

International Conference on Document Analysis and Recognition (ICDAR)
International Journal of Document Analysis and Recognition (IJDAR)
Document Image Analysis for Libraries (DIAL)
International Conference on Pattern Recognition (ICPR)
IEEE Transactions on Pattern Analysis and Machine Intelligence (PAMI)
International Workshop on the Frontiers of Handwriting Recognition (IFWHR)
Association for Image and Information Management (AIIM)

What Makes Document Recognition Difficult?

- variability

- 100000s of journal, magazine, newspaper formats

- unpredictability, exceptions

- handwritten annotations, post-it notes, strange fonts, languages
- figures, graphics can include anything

- context

- meaning of glyphs, graphics, whitespace depend on---and define---genre and style

- problem definition

simple transcription ← ? → build a resource for exploring knowledge

- You can engineer a solution to any narrowly constrained problem.

- Real applications seldom obey narrow problem constraints.