



Document Image Processing for Balinese Palm Leaf Manuscripts *Protocol, Scheme, and Challenges*

Made Windu Antara Kesiman, S.T., M.Sc., Ph.D

antara.kesiman@undiksha.ac.id — research.undiksha.ac.id/vvip-rg

Virtual, Vision, Image, and Pattern Research Group
Department of Informatics Engineering
Faculty of Engineering and Vocational
Universitas Pendidikan Ganesha

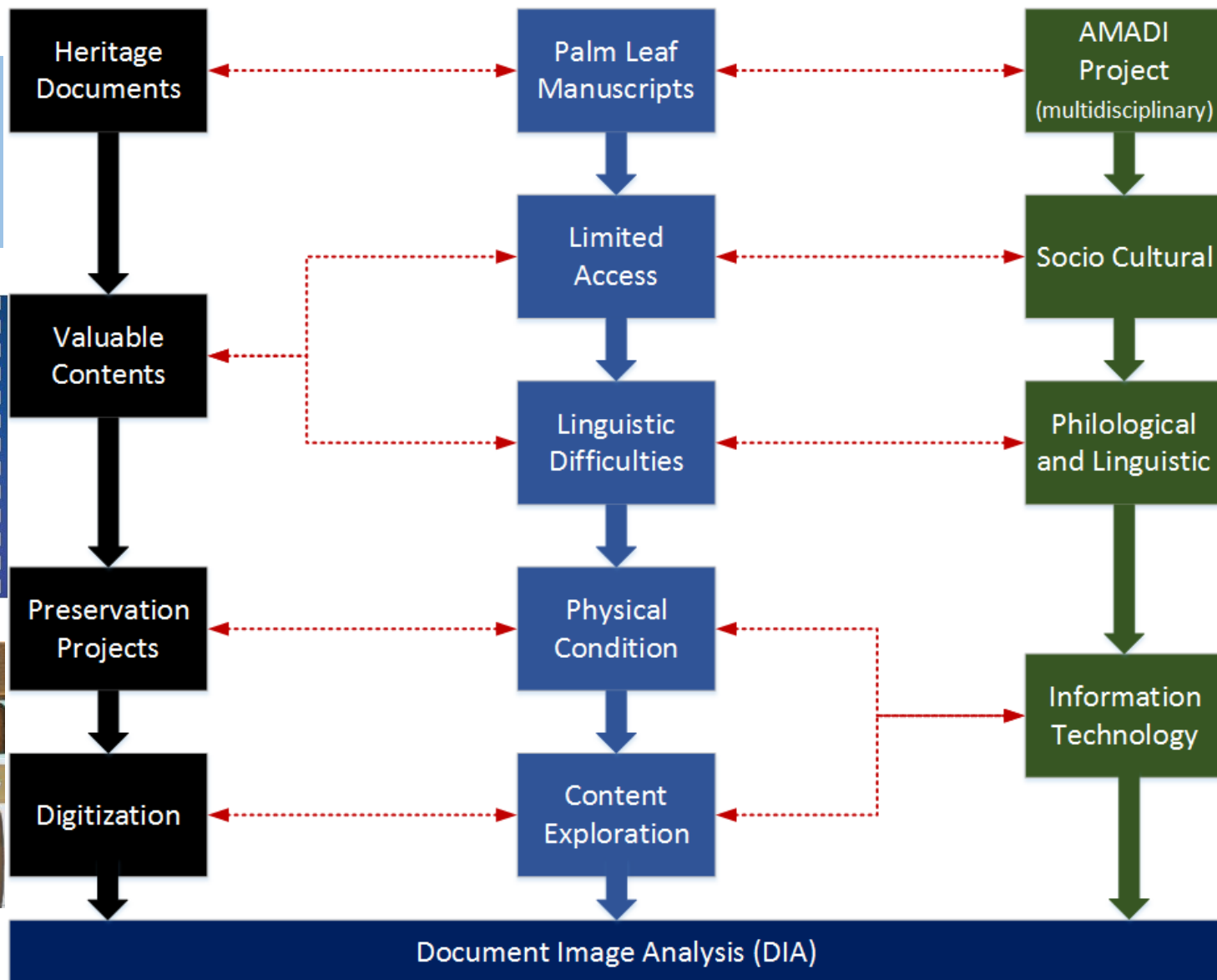


Outline

- Research Contexts :
 - ☐ Projects
 - ☐ Motivations
 - ☐ Objectives
- Research Problematics
 - ✓ Challenges
 - ✓ Protocol
 - ✓ Scheme



Research Contexts



Project :: *from AMADI Project to EpsiLont Project*

2014 – 2018



2019-2021-??

AMADI Project

Ancient Manuscripts Digitization and Indexation

STIC Asia Program - French Ministry of Foreign Affairs and International Development (MAEDI).

Partners: France, Indonesia and Cambodia



- Laboratory Informatique Image et Interaction (L3i) of ULR, France
- Laboratory Cultural Informatics (LCI) of UNDIKSHA, Bali, Indonesia
- Laboratory RAID and Center for Sundanese Culture Studies of UNPAD, Bandung, Indonesia
- Laboratory GIC of ITC, Phnom Penh, Cambodia

EpsiLont Project

Electronic Pattern Analysis for Lontar

- PDUPT Ristekbrin 2019-2021
- Penelitian Terapan DIPA Undiksha 2020

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Balinese LONTAR

Collections

Museum Gedong Kertya and Bali



± 6,000 collections

religion, holy formulae, rituals, family genealogies, law codes, treaties on medicine (usadha), arts and architecture, calendar, prose, poem and even magic

Private Family



± 50,000 collections

Writing and Materials

Palm tree



small pen-knife *pengerupak*



scratching palm leaves



rubbed with natural black dye



cleaned up with cotton



Transliteration Books (in Indonesian language : Alih Aksara)

- hard to associate the collection
- belongs to a collection which is not stored in the Museum

Catalog (example): Lontar Collection Museum Bali, Denpasar

Rak I.1

BABAD

1. Babad Brahma Siwa - 07.252/6103
2. Prasasti Ngurah Sidemen - 07.262/5353
3. Mpu Kuturan - 07.256/6162
4. Babad Sukawati - 07.266/6291
5. Babad Bali Pulina - 07.270/6149
6. Babad Brahmana Manuaba - 07.251/6150
7. Prasasti Buleleng - 07.164/6153
8. Usana Bali - 07.269/6289
9. Brahmangsa - 07.158/6152
10. Babad Mayadanawa - 07.257/5381
11. Babad Ularan - 07.267/6363/N
12. Pamancangah Dalem - 07.259/5165/IV.a
13. Babad Mangwi - 07.258/6366
14. Prasasti Pande Capung - 07.260/6361
15. Babad Danghyang Nirarta - 07.255/5971
16. Babad Pasek - 07.268/5951
17. Babad Calonarang - 07.203/6596/N
18. Pamancangah Badung - 07.209/5290
19. Prasasti Dukuh Kedangan - 07.24/01/MB
20. Babad Brahma Cute - 07.250/6362
21. Babad Bandesa Sakti Beng - 07.248/6364
22. Babad Sangging - 07.264/6367

Rak I.2

GAGURITAN

1. Peparikan Lawe - 07.308/6151
2. Peparikan Tantri - 07.278/5164.3
3. Gaguritan Jayaprana - 07.282/6105/N
4. Gaguritan Aji Tatwa Mimit - 07.284/5275
5. Gaguritan Gunawati - 07.280/6108
6. Gaguritan Sasak - 07.283/5284
7. Gaguritan Dewa Manggis - 07.281/5310
8. Gaguritan Sampik - 07.213/3813.3

Rak II.2

MANTRA ASTAWA

1. Utama Japa - 07.158/6470
2. Puja Srawa - 07.145/05/MB
3. Panyapa Brahma Wisnu - 07.131/5958
4. Kusuma Dewa - 07.33/5102
5. Tata Kramaning Pabaktian - 07.299/6467
6. Ludra Pinggala - 07.148/5956
7. Puja Serawa - 07.144/6072
8. Banyu Awang - 07.107/6589
9. Puja Soha - 07.129/6587
10. Pangaskara Wong Mati - 07.128/5278/I.c
11. Pangembak Mantra - 07.146/5687

Rak I.3

KAKAWIN

1. Kakawin Bharatyuddha - 07.17/6058
2. Bharatyuddha Grantang Basa - 07.19/5862
3. Kakawin Bharatyuddha - 07.16/5635.3
4. Bharatyuddha Kreta Basa - 07.18/5832
5. Kakawin Arjuna Sastrabahu - 07.6/6165
6. Kakawin Harisraya - 07.29/5157.3
7. kakawin Kangsa - 07.32/5830
8. Kakawin Bhomantaka - 07.249/5859
9. Kakawin Sutasoma - 07.48/5858
10. Arjunawiwaha Grantang Basa - 07.7/6030

Rak I.6

KIDUNG

12. Kidung Tantri Kamandaka - 07.276/5699
13. Kidung Lawe - 07.60/5052
14. Kidung Pararaton - 07.62/5053
15. Kidung Tantri Kamandaka - 07.275/5545
16. Kidung Prembon Edan Wirangrong - 07.64/5035
17. Kidung Suphala Sidanta - 07.67/6679
18. Kidung Dhamar Wulan - 07.274/5160

Rak I.4

KAKAWIN

11. Arjuna Wiwaha - 07.8/5621
12. Arjuna Wiwaha - 07.9/5944
13. Kakawin Bharatayuddha - 07.34.A/5099
14. Kakawin Ariwangsa - 07.5/5547
15. Bharatayuddha Grantang Basa - 07.34/3781
16. Arjuna Wiwaha Grantang Basa - 07.10/5863
17. Kakawin Ramayana - 07.40/5634.3
18. Kakawin Ramayana - 07.41/5634.3
19. Bharatayuddha Grantang Basa - 07.20/5941
20. Arjuna Wiwaha - 07.11/5864
21. Kakawin Ramayana - 07.42

Rak II.1

KIDUNG

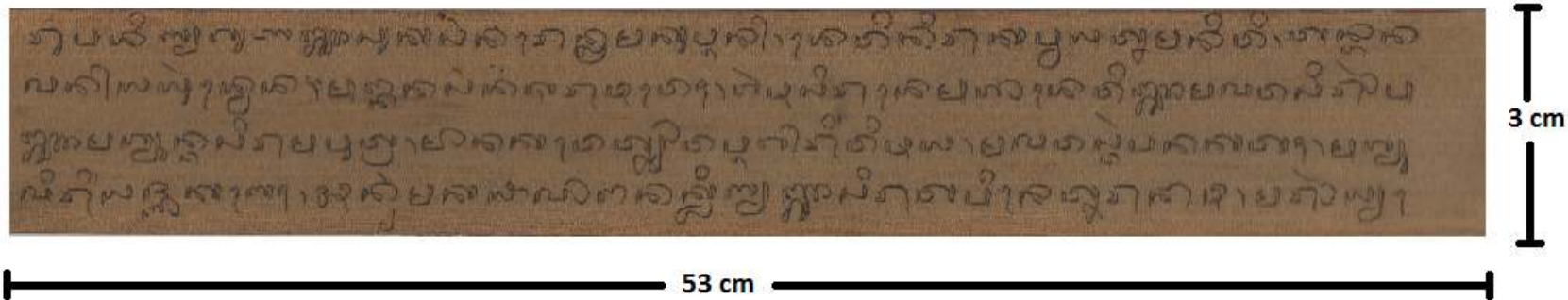
1. Kidung Rusak Sasak - 07.66/5056
2. Kidung Aji Pasurwan - 07.56/5054
3. Kidung Wijaya Krama - 07.69/5103
4. Kidung Widhari Smara - 07.51/5562
5. Kidung Bagus Umbara - 07.57/5351.2
6. Kidung Cupak - 07.58/5057
7. Kidung Tantri - 07.279/5062.3
8. Kidung Tantri - 07.247/6029
9. Kidung Amad Raden Saputra - 07.279/3814
10. Kidung Arsawijaya - 07.55/5058.3
11. Kidung Panji - 07.142/6062

Rak I.5

KAKAWIN

22. Kakawin Bharatayuddha - 07.15/5826
23. Kakawin Ramayana - 07.38/5059/IV/b
24. Kakawin Ramayana - 07.17/5970
25. Kakawin Bhuta Yadnya - 07.24/5822
26. Kakawin Sang Hyang Kala - 07.44/6598

Physical Dimension of Lontar



- Normally, in one leaf, there are 4 text lines
- Number of leaf for a collection, example :
 - ✓ For the collection of **Kakawin Ramayana - 07.41/5634.3** = 140 leafs
 - ✓ For the collection of **Kakawin Ramayana - 07.40/5634.3** = 155 leafs



Production Process of Lontar

✓ Sidemen Village, Region of Karangasem, Bali



Production Process of Lontar



Choose the leaves: a little old,
wide enough, smooth and flat



Production Process of Lontar



Soak in cold water (for 10 days) - the leaves will become a bit soft

Production Process of Lontar



Boil for 4 hours + ingredients and spices e.g. tea and pepper to prevent bugs

Production Process of Lontar



Dry in the sun for 2/3 days

Production Process of Lontar

Define the size and cut



Make three hole markers



Tighten and press with wood



Production Process of Lontar



Smoothing the edges of the leaves

Production Process of Lontar



Create lines with carbon to guide the writing



Production Process of Lontar



Tighten and press with wood

Production Process of Lontar

Ready to write



Writing Process and the Script

Script

- The writings:
 - the ancient literary texts composed in the old Javanese of Kawi and Sanskrit;



<http://www.wonderfullbali.com>

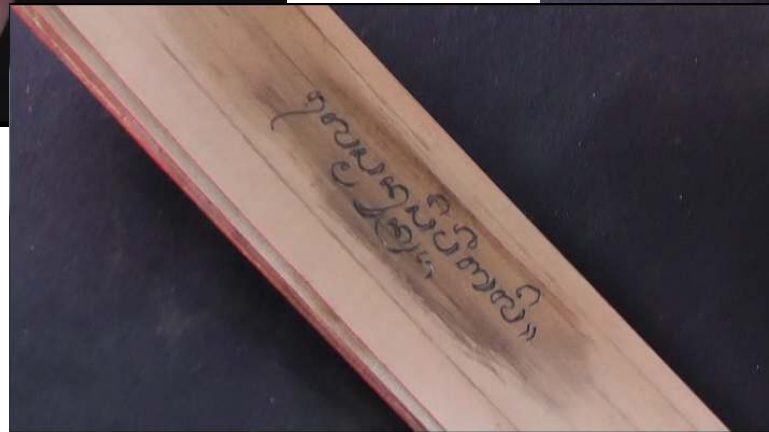
Lontar are inscribed with a special tool called a pengerupak. It is made up of iron, with its sharp tip in a triangular shape so it can make inscriptions thick and thin. There are two types of pengerupak, one for writing and one for drawing. The pengerupak for writing is about 15 in length and 1.5 cm wide, the pengerupak for drawing is the same length but is only 0.5 to 1 cm wide. There is also a third long type which is used for cutting rontar leaves.

Writing Process

Lontar are inscribed with a special tool called a pengerupak. It is made up of iron, with its sharp tip in a triangular shape so it can make inscriptions thick and thin.



Ancient palm leaf manuscript is written on a dried palm leaf using a kind of small knife, which is then rubbed with natural dyes



The writings were incised on one (and or both) sides of the sheet with a sharp knife and the script is therefore blackened with soot





The Unicode Standard, Version 7.0

Balinese, Range : 1B00 – 1B7F

Various signs

1B00		BALINESE SIGN ULU RICEM = ardhaçandra
1B01		BALINESE SIGN ULU CANDRA = candrabindu
1B02		BALINESE SIGN CECEK = anusvara
1B03		BALINESE SIGN SURANG = repha
1B04		BALINESE SIGN BISAH = visarga

	1B0	1B1	1B2	1B3	1B4	1B5	1B6	1B7
0	◌် 1B00	◌ျ 1B10	◌ဲ 1B20	◌ဲ 1B30	◌ံ 1B40	◌ံ 1B50	◌ံ 1B60	◌ံ 1B70
1	◌ံ 1B01	◌ံ 1B11	◌ံ 1B21	◌ံ 1B31	◌ံ 1B41	◌ံ 1B51	◌ံ 1B61	◌ံ 1B71
2	◌ံ 1B02	◌ံ 1B12	◌ံ 1B22	◌ံ 1B32	◌ံ 1B42	◌ံ 1B52	◌ံ 1B62	◌ံ 1B72
3	◌ံ 1B03	◌ံ 1B13	◌ံ 1B23	◌ံ 1B33	◌ံ 1B43	◌ံ 1B53	◌ံ 1B63	◌ံ 1B73
4	◌ံ 1B04	◌ံ 1B14	◌ံ 1B24	◌ံ 1B34	◌ံ 1B44	◌ံ 1B54	◌ံ 1B64	◌ံ 1B74
5	◌ံ 1B05	◌ံ 1B15	◌ံ 1B25	◌ံ 1B35	◌ံ 1B45	◌ံ 1B55	◌ံ 1B65	◌ံ 1B75

6								
	1B06	1B16	1B26	1B36	1B46	1B56	1B66	1B76
								
7								
	1B07	1B17	1B27	1B37	1B47	1B57	1B67	1B77
								
8								
	1B08	1B18	1B28	1B38	1B48	1B58	1B68	1B78
								
9								
	1B09	1B19	1B29	1B39	1B49	1B59	1B69	1B79
								
A								
	1B0A	1B1A	1B2A	1B3A	1B4A	1B5A	1B6A	1B7A
B								
	1B0B	1B1B	1B2B	1B3B	1B4B	1B5B	1B6B	1B7B

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Motivations and Objectives

Motivations

- to bring added value to digitized palm leaf manuscripts
 - analyze
 - transliterate
 - index



more accessible, *readable* and *understandable*

... in working with the collections of palm leaf manuscripts

- **preservation**
 - physical condition of material
- **access**
 - very valuable cultural content
- **sharing**
 - open to community

Technical Objectives

to develop a DIA system for document images of palm leaf manuscripts

Research Problematics :: *Very Wide Range ! Pipeline!!*

- ✓ **Digitization**>
- ✓ **Ground Truth Dataset Construction**>
- ✓ Document Image Pre-processing>
- ✓ **Binarization**>
- ✓ **Segmentation/Localization/Detection**>
 - ✓ Physical : **Text Area**, Paragraph, Figure/Diagram/Logo/Letterine/**Prasi Lontar**, **Textline**, Word, **Character/Glyph**, Tabel
 - ✓ Semantic : Header/Footer, Footnote, Formula/Equation, Title, Comic components
- ✓ **Recognition**>
 - ✓ Text : **Textline**, **Word**, **Character/Glyph**, Formula/Equation
 - ✓ Image : Figure/Diagram/Logo/Letterine, Shape and Curve, **Prasi Lontar**
- ✓ Identification : Script, Writer>
- ✓ **Transliteration + Post Transliteration Correction**>
- ✓ **Word Spotting**>
- ✓ **Document Indexation/Retrieval**>

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Challenges

Socio-cultural Challenges

Difficulty in collecting samples

- cultural and religious conditions
- sacred collections
- permission
- rules
- families are very reluctant

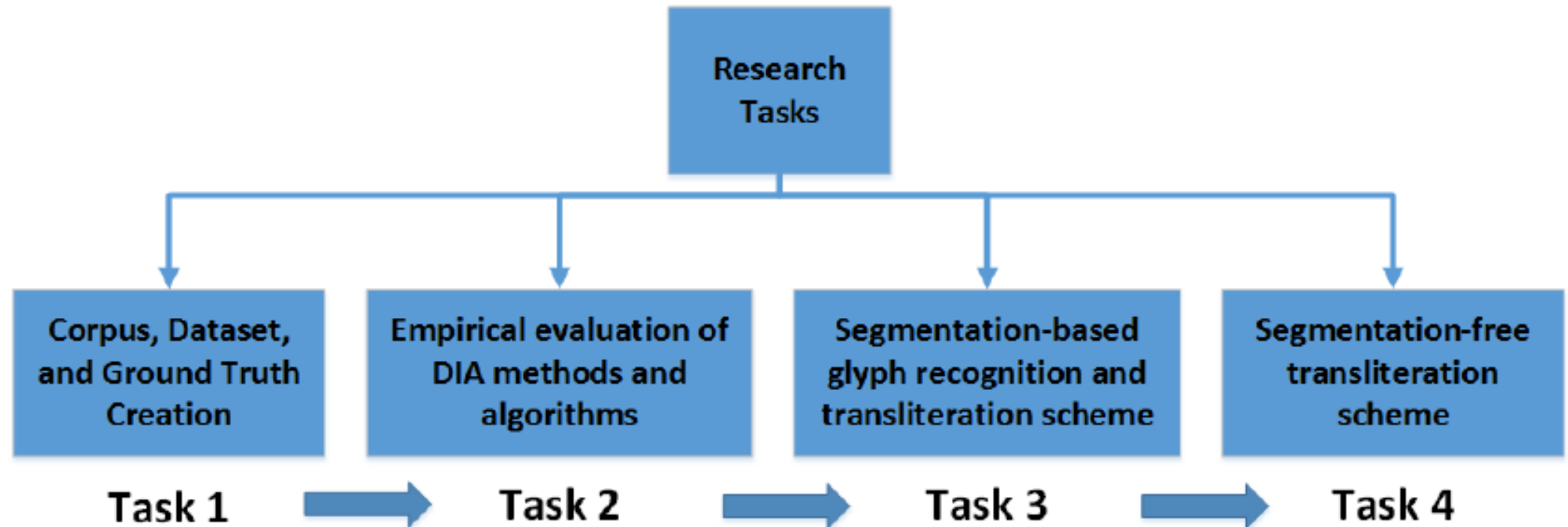


Difficulty in finding the Balinese philologist

- not many Balinese can read
- not used everyday
- not popular
- retirement
- lack experience



Global Overview



Research Tasks

Corpus, Dataset,
and Ground Truth
Creation

Empirical evaluation of
DIA methods and
algorithms

Segmentation-based
glyph recognition and
transliteration scheme

Segmentation-free
transliteration
scheme

Task 1

Task 2

Task 3

Task 4

- 1.1 Collecting and digitizing the manuscript corpus
- 1.2 Designing overall scheme of ground truth construction
- 1.3 Proposing specific scheme for the construction of the ground truth of binarized images
- 1.4 Analyzing the subjectivity of the human intervention and ground truth variability
- 1.5 Presenting the AMADI_LontarSet

Challenges :: *Digitization*



- ✓ Size of Lontar
- ✓ Lontar pages packaging
- ✓ Fragility of Lontar material
- ✓ Sacred collection and location

Solution :: *Digitization* :: *Corpus Collection*

23 different collections

- 5 different locations (regions):
 - 2 museums
 - 3 private families



several restricted conditions

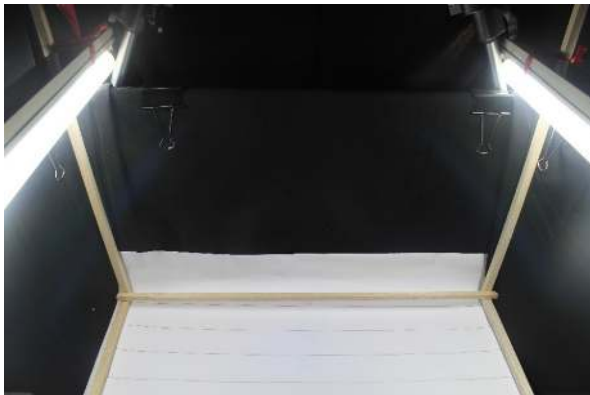
Camera Support



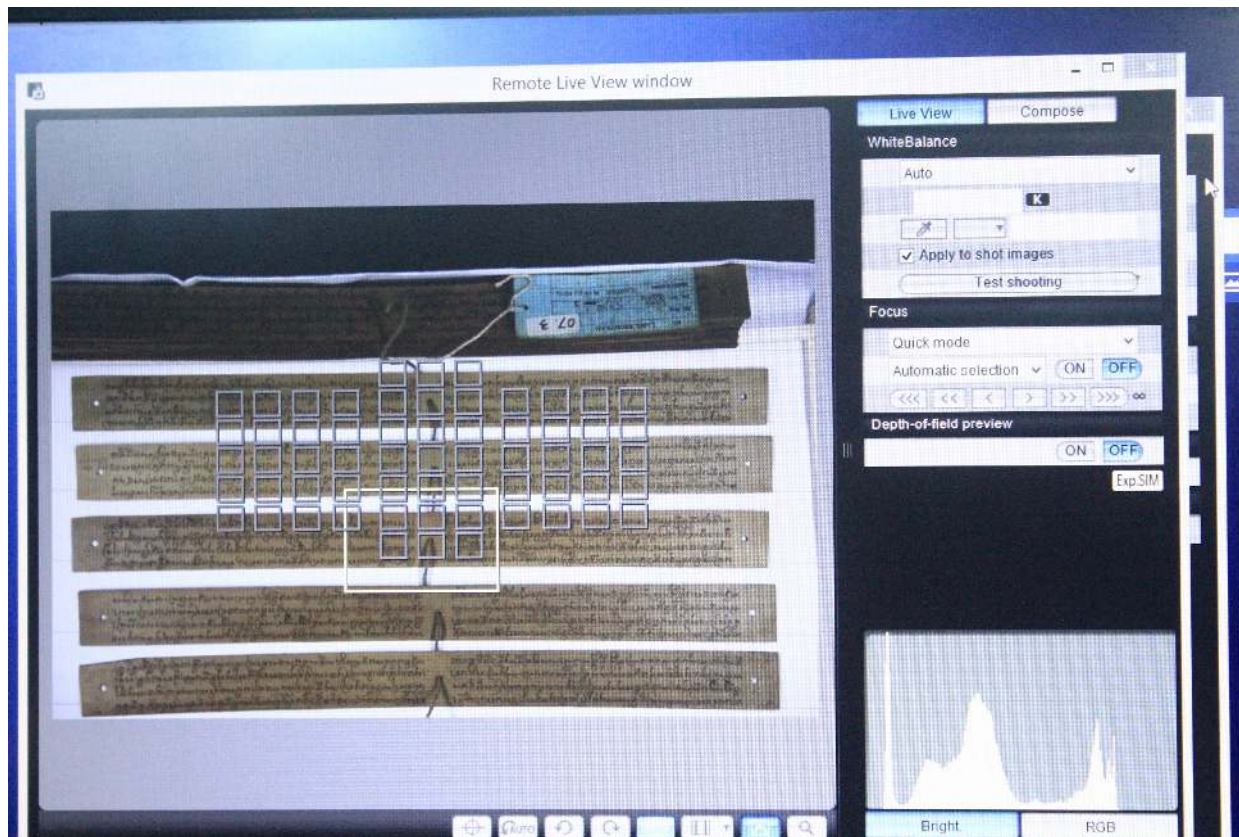
Digitization Process



Solution :: *Digitization Support*



Solution :: *Digitization :: Image Standardization*



Camera model : Canon EOS 5D Mark III
F-stop : f/22 (diafragma)
Exposure time : 1/50 sec
ISO speed : ISO-6400
Focal length : 70 mm
Flash : On - 1/64
Additional light : White Neon 50 cm 20 watt
Distance to object : 76 cm
Length object (manuscript) : +/- 50 cm
Focus : Quick mode - Auto selection On
Sensor size : 36x24 mm

Solution :: *Digitization Process*



Solution :: *Digitization Process :: Image Samples*

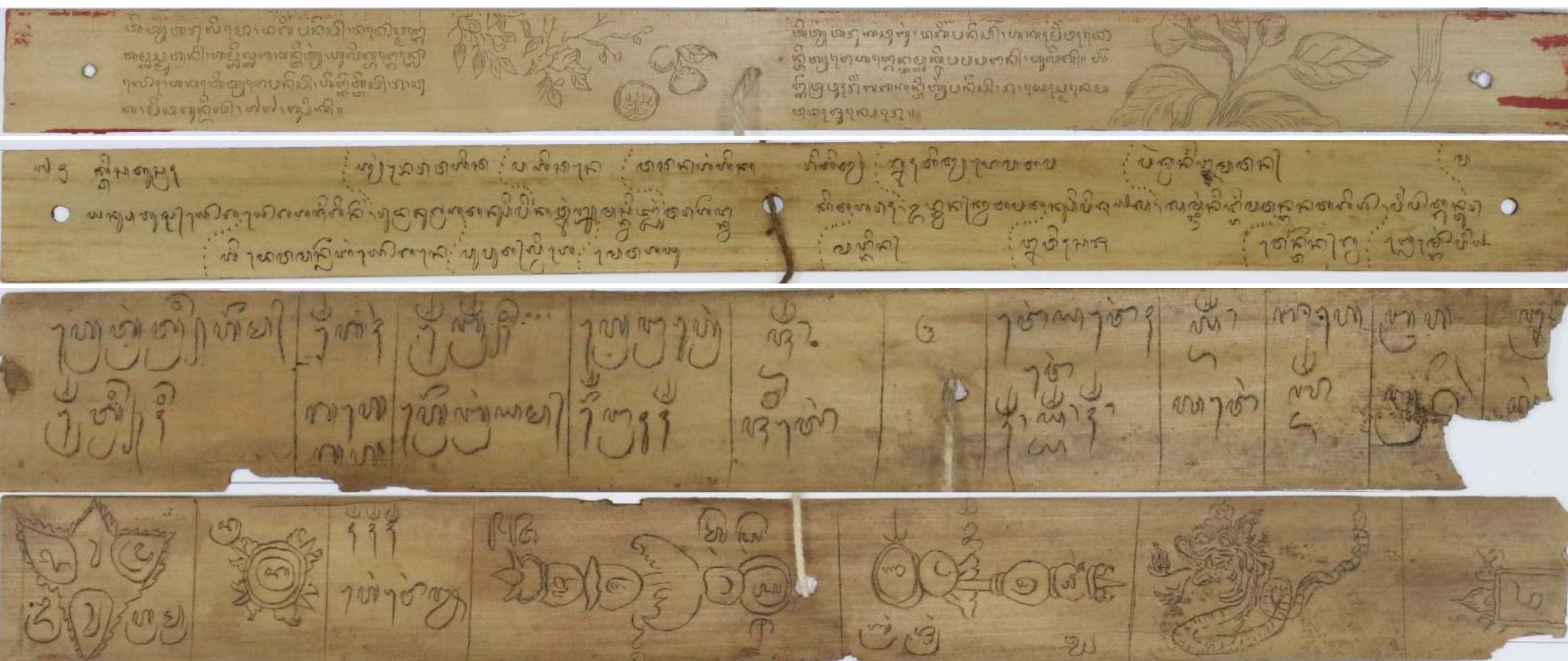


Sample Images of Manuscript from Museum Bali

[illegible]

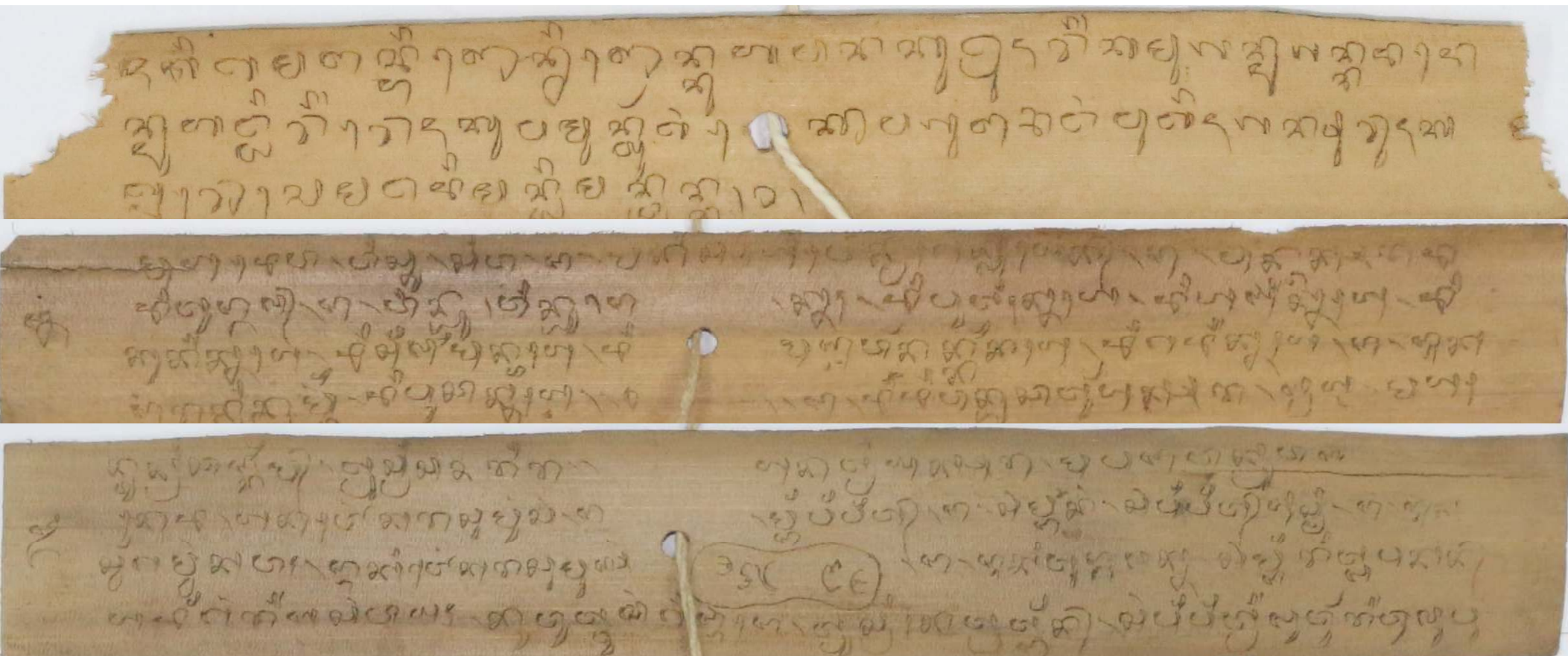
Protocol 1 :

Sample Images of Manuscript from Jagaraga



Protocol 1 :

Sample Images of Manuscript from Jagaraga



Protocol 1 :

Sample Images of Manuscript from Bangli



Protocol 1 :

Sample Images of Manuscript from Rendang



Summary Corpus Collection

Location	NbCollection	Collection	NbPages
Collection Privee, Keluarga Bangli, Bangli	1	Bangli	82
Museum Gedong Kertya, Singaraja	10	IIA-10-1534	8
		IIA-5-789	8
		IIB-2-180	8
		IIIB-12-306	8
		IIIB-42-1526	8
		IIIB-45-2296	8
		IIIC-19-1293	8
		IIIC-20-1397	8
		IIIC-23-1506	8
		IIIC-24-1641	8
Collection Privee, Keluarga Jagaraga, Buleleng	7	JG-01	16
		JG-02	10
		JG-03	16
		JG-04	12
		JG-05	8
		JG-06	5
		JG-07	10
Museum Bali, Denpasar	4	MB-AdiParwa(Purana)-5338.2-IV.a	40
		MB-AjiGriguh-5783-107.2	20
		MB-ArjunaWiwaha-GrantangBasall	30
		MB-TaruPramana	40
Collection Privee, Keluarga Rendang, Karangasem	1	WN	24
Total	23		393

Protocol 2 : Transliteration of Manuscripts



Bangli-P39

- Buik julang, ijo asuh, ta. Biing kuning rarajah soring tegil, ja, biing putih rarajah, ta. Biing kuning raraja
- H jarijinia, ja, nga, biing nyelem, ta. Biing jenar, ta. Biing jenar rarajah, ja, ijo jenar rarajah tuk muhmuh, tekaning putih
- Tulus, sabiru, wangkas biru, klau biru, ta. Biinga ijo, ja, putih tulus, putih siungan
- Ta. Biing bang karna, ja, buik kuning, ta. Biing gadang cucuk cemeng, ja, srawah putih, ta. Biing nyelem biing gadang

Protocol 2 : Transliteration of Manuscripts



IIIB-45-2296-P7

- Nto cai rese pang apang pasti, di keneh yen suba anut ban cai ngonek sastra, pasang sutra keras ento masih tuptup, kalih munyining pidartane madan pidarta munyi. Sastra soroh matunggalan,
- Munyi tunggal mawa ada wuwuhin, nyan pidarta sangkan payu, dadi anto kaucap, di madune da dantia munyine patuh, ada len buin kaucap, sa dantia munyinya tunggal. Lan ta
- Lawia samurda, sok len sambat masih mamunyi tunggal, ta lati ta dantia patuh, masih teke talawia. Ada len to macelek suku kembang masih mawak talawia, tunggal munyinya wia
- Kti. Puh Durma. Ga gora ga dantia, mamunyi tunggal, na dantiane tekening ne madan na rambat, tunggal masih munyinya pe palane tunggal munyi teken pa dantia,



MB-TaruPramana-P4

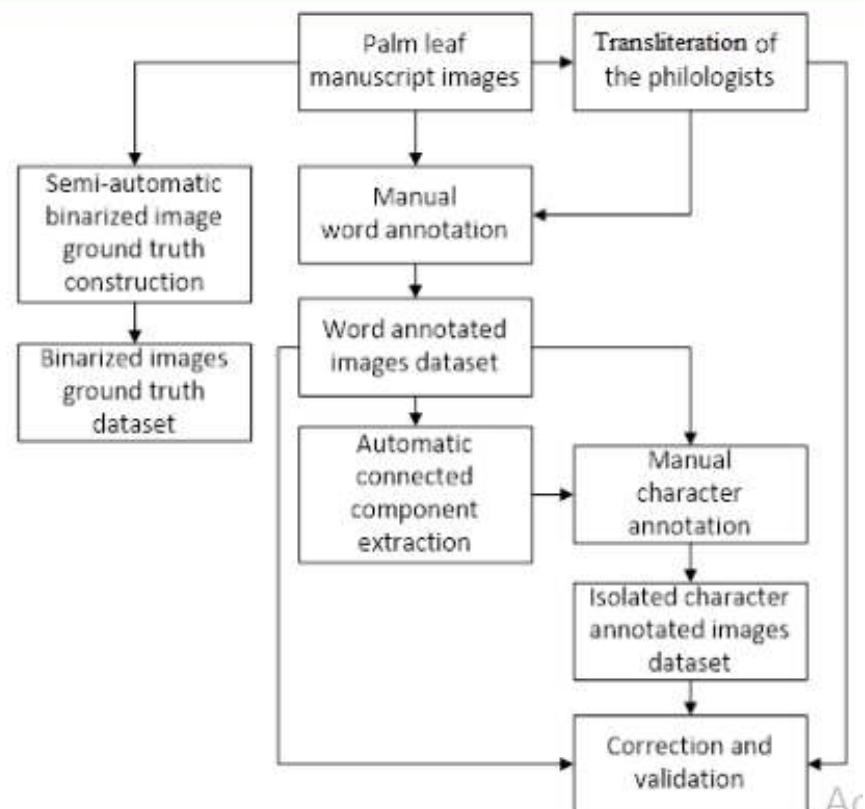
- Kadi titiang. Ngandika sang prabhu. Kene iba binglin, wireh awake dadi baleen, tan sida nyegerang jani,
- Makeneh nyekenang teken I kayu niri niri. Apa gunannyane, muang dadi anggon uba
- D apa. I taru binglin mapamit. Sang prabhu empu kuturan malih ngarad. Raris rauh wit kasela gui, sasampu
- Ne tinakenan annuli matur. Titiang mawasta kasela gui, daging titiange tis, dadong anggen tamba lolo

Scheme :: *Ground Truth Dataset Construction*

Approach

- man-hours
- expertized people
 - manual
 - semi automatic
- ground truther
 - Balinese philologist
 - students in Balinese literature
 - students in Informatics

Overall Scheme



Protocol :: *Ground Truth Dataset Construction*

Transliterated Text - Manual



III B-45-2296-P7

- Nito cai resempang apang pasti, di keneh yen suba anut ban cai ngonek sastra, pasang sul
- Munyi tunggal mawa ada wuwuhin, nyan pidarta sangkan payu, dadi anto kaucap, di ma
- Lawia samurda, sok len sambat masih mamunyi tunggal, ta lati ta dantia patuh, masih te
- Kti. Puh Durma. Ga gora ga dantia, mamunyi tunggal, na dantiane tekening ne madan ni

Text Line Segmented - Manual (from Binarized Image)



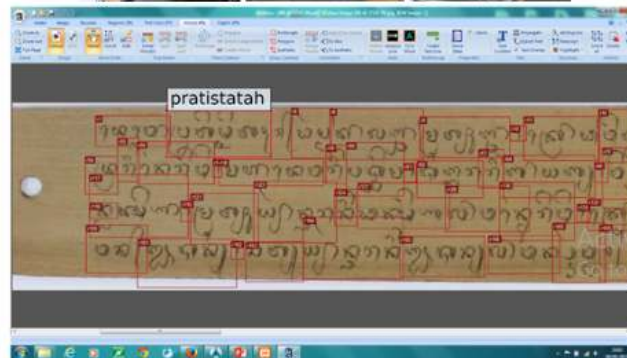
Binarized Image - Semi Auto

Proposed Binarization Scheme

Glyph Annotated - Semi Auto

[illegible]

Word Annotated - Manual



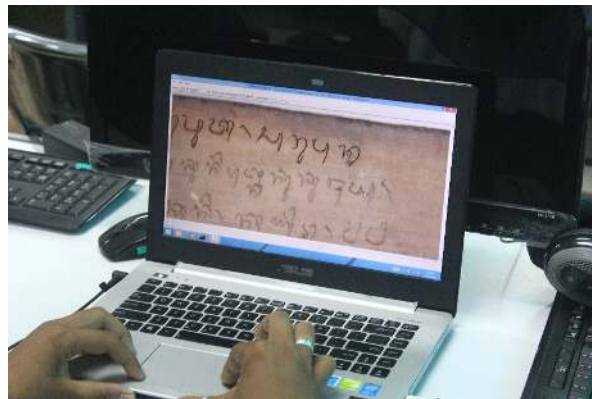
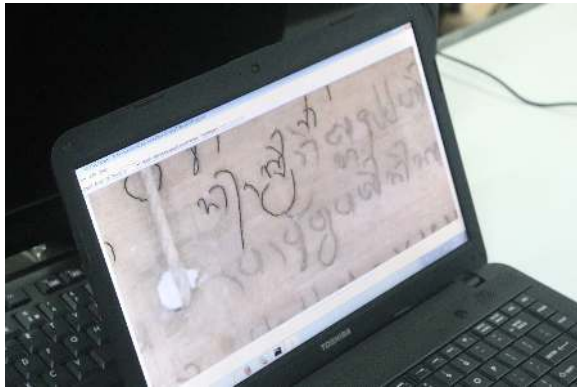
Ground Truthers:

- ✓ 70 students of Informatics
- ✓ 10 students of Balinese Litterature
- ✓ 2 Balinese philologists

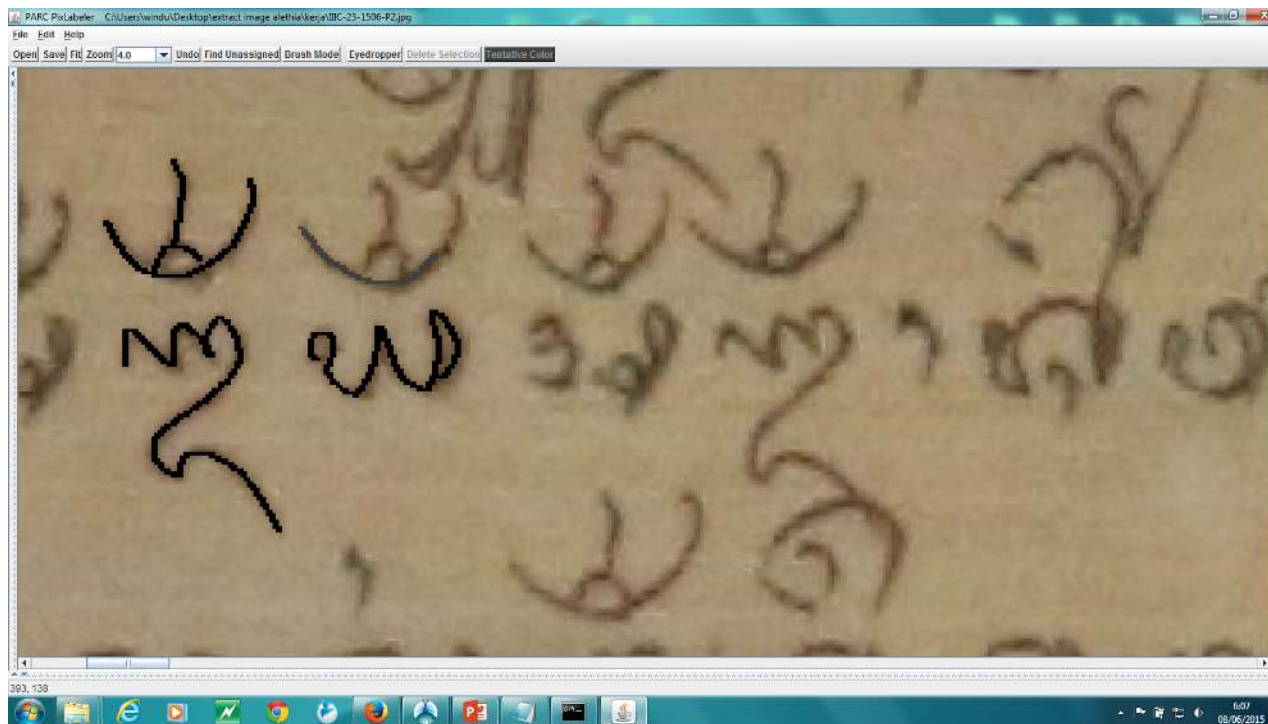
Tools:

- ✓ **PixLabeler** for Binarization
- ✓ **Aletheia** for Word Annotation
- ✓ Developed a **Web-based interface** for Glyph Annotation
- ✓ **Text Editor** for Text Transliteration

Process :: *Ground Truth Dataset Construction*



Protocol 3 : Ground Truth Creation Binarized Image Dataset



Tool : PixLabeler

[E. Saund, J. Lin, and P. Sarkar, "PixLabeler: User Interface for Pixel-Level Labeling of Elements in Document Images," presented at the ICDAR '09. 10th International Conference on Document Analysis and Recognition, 2009, pp. 646–650]

Result : Estimated Ground Truth of A Nondegraded Palm Leaf Manuscript Image

[illegible]

Result : Estimated Ground Truth of A Degraded Palm Leaf Manuscript Image

[illegible]

Skeletonized Ground Truth – Manual Corrected

[illegible]

Ground truth image constructed without any constraint of initial binarized image

Result : Estimated Ground Truth of A Degraded Palm Leaf Manuscript Image

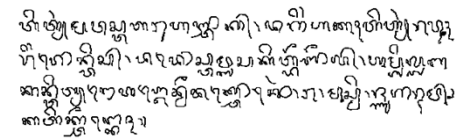
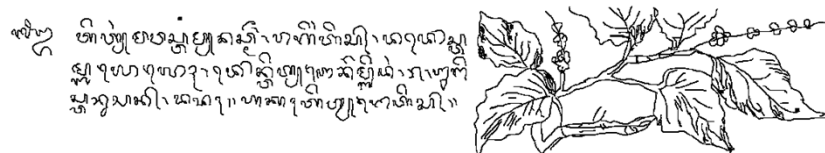
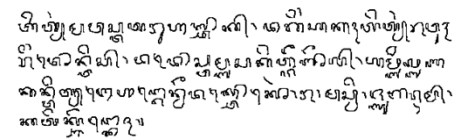
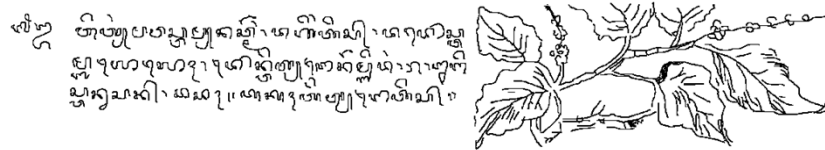


အဲဒီပုဒ်ကလေးဟာ အလွန်အမင်း
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အလွန်အမင်းအလွန်အမင်း



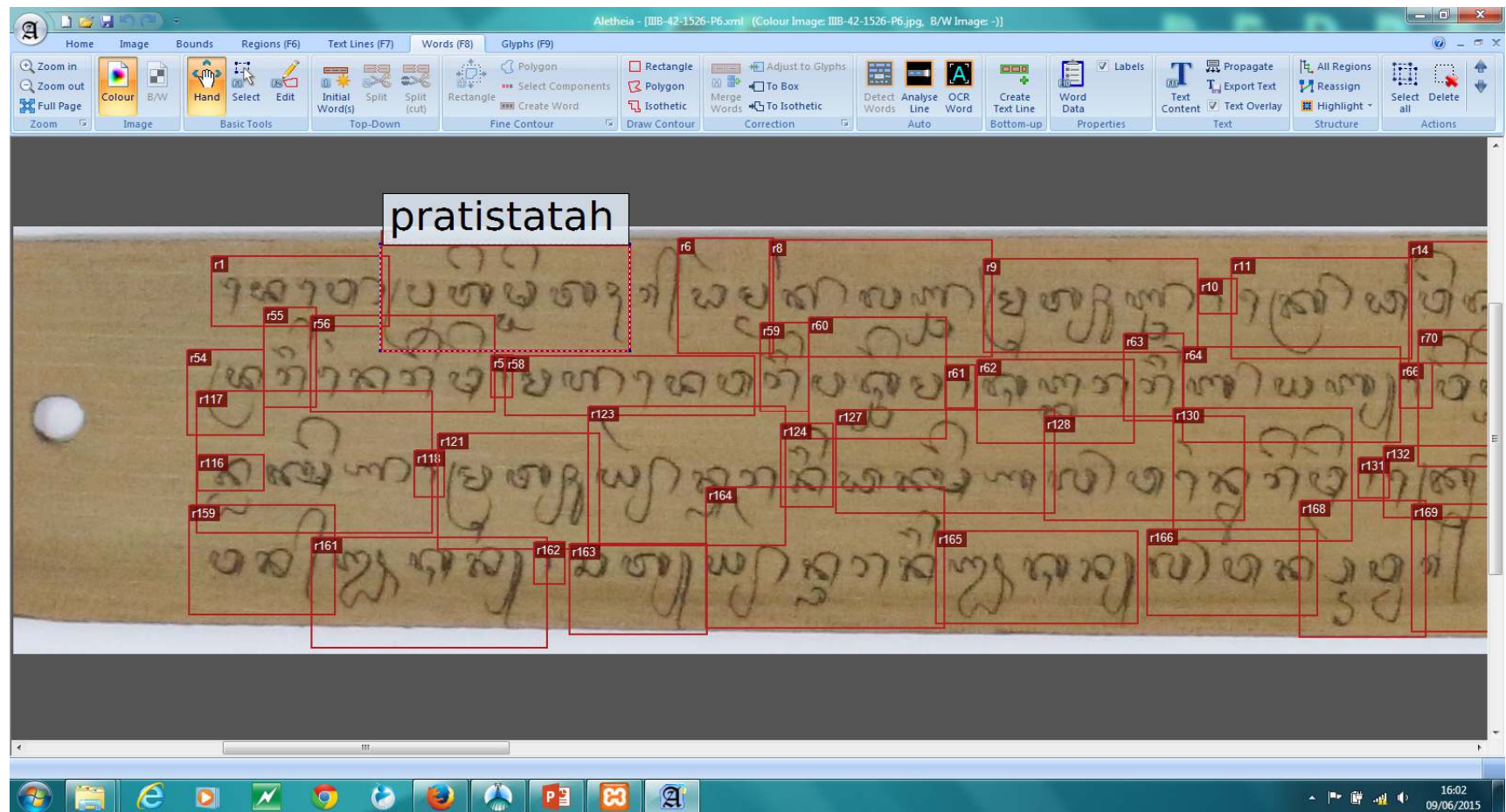
အဲဒီပုဒ်ကလေးဟာ အလွန်အမင်း
အလွန်အမင်းအလွန်အမင်း
အလွန်အမင်းအလွန်အမင်း
အလွန်အမင်းအလွန်အမင်း

Ground Truth Binarized Image Variability

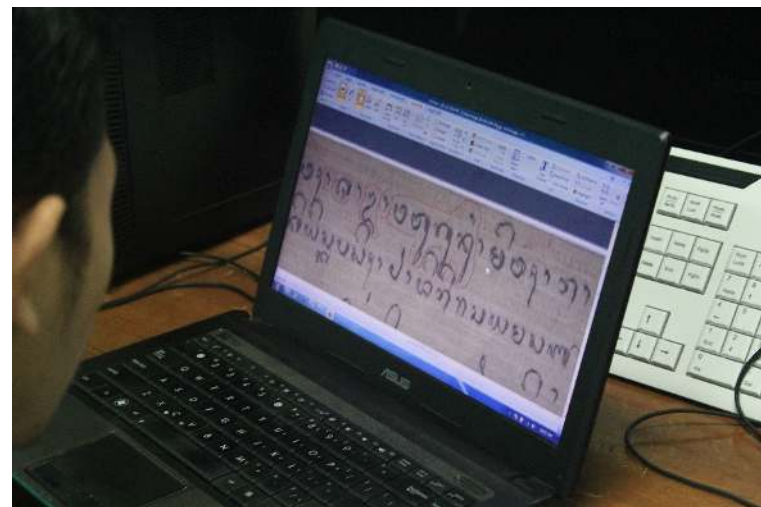


Protocol 4 : Ground Truth Creation

Word Annotated Image Dataset



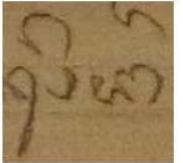
Protocol 4: Ground Truth Creation Word Annotated Image Dataset



Data: Word Annotated Image Dataset

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    <Creator></Creator>
    <Created>2015-06-01T02:14:47</Created>
    <LastChange>2015-06-01T02:14:47</LastChange></Metadata>
    <Page imageFilename="IIIB-42-1526-P6.jpg" imageWidth="5755" imageHeight="477">
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        <Coords points="0,0 1,0 1,1 0,1"/>
        <TextLine id="tempLine357564684568544579089">
          <Coords points="0,0 1,0 1,1 0,1"/>
          <Word id="r1">
            <Coords points="212,34 402,34 402,112 212,112"/>
            <TextEquiv>
              <Unicode>dewo</Unicode></TextEquiv></Word>
            <Word id="r3">
              <Coords points="393,21 659,21 659,139 393,139"/>
              <TextEquiv>
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            <Word id="r6">
              <Coords points="711,14 813,14 813,142 711,142"/>
              <TextEquiv>
                <Unicode>Dharma</Unicode></TextEquiv></Word>
            <Word id="r8">
              <Coords points="809,16 1047,16 1047,146 809,146"/>
              <TextEquiv>
                <Unicode>kanglancang</Unicode></TextEquiv></Word>
            <Word id="r9">
              <Coords points="1038,37 1267,37 1267,141 1038,141"/>
              <TextEquiv>
                <Unicode>mretyuanca</Unicode></TextEquiv></Word>
            <Word id="r10">
              <Coords points="1268,59 1309,59 1309,98 1268,98"/>
              <TextEquiv>
                <Unicode>,</Unicode></TextEquiv></Word>
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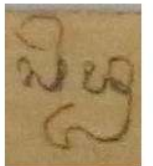
sida_IIA-10-1534-P6_r52_1281-110-1418-239.jpg



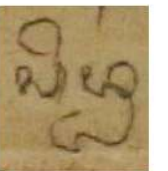
sida_IIA-10-1534-P7_r9_472-0-613-118.jpg



sida_IIIC-19-1293-P7_r5_1274-11-1383-143.jpg



sida_IIIC-19-1293-P7_r127_4837-333-4958-466.jpg

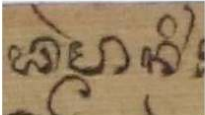

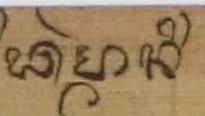

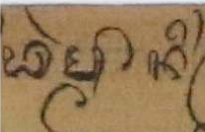

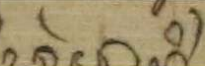



Protocol 5 : Ground Truth Creation Character Annotated Image Dataset

1	dharmaning_MB-AdiParwa(Purana)-5338.2-IV.a-P10_r4_502-28-707-145.jpg		      <input type="button" value="Simpan Segmen"/>
2	dharmaning_MB-AdiParwa(Purana)-5338.2-IV.a-P10_r45_2252-8-2464-127.jpg		     <input type="button" value="Simpan Segmen"/>

Web Dataset for Palm Leaf Manuscript Images

Screenshot of a web application interface for a Palm Leaf Manuscript Image Dataset. The browser shows multiple tabs, including 'List Image Word : dharm...', and the address bar displays 'localhost/prot/show_img.php?word=dharmaning'.

No	Filename	Image	Segmen Huruf
1	dharmaning_MB-AdiParwa(Purana)-5338.2-IV.a-P10_r4_502-28-707-145.jpg		 DHA MA SURANK GANTUI MA TED NI CECEK
2	dharmaning_MB-AdiParwa(Purana)-5338.2-IV.a-P10_r45_2252-8-2464-127.jpg		 DHA MA SURANK GANTUI MA TED NING
3	dharmaning_MB-AdiParwa(Purana)-5338.2-IV.a-P10_r72_3009-5-3218-145.jpg		 DHA MA SURANK GANTUI MA TED NI CECEK
4	dharmaning_MB-AdiParwa(Purana)-5338.2-IV.a-P10_r72_3009-5-3218-145.jpg		 DHA MA SURANK GANTUI MA TED NI CECEK

The interface includes a search bar labeled 'Rechercher' and a 'Simpan Segmen' button for each row. The Windows taskbar at the bottom shows the date and time as 5:30 on 15/01/2016.

AMADI_LontarSet :

The first handwritten balinese Palm Leaf Manuscripts dataset

- **ICFHR 2016 Competition on the Analysis of Handwritten Text in Images of Balinese Palm Leaf Manuscripts**
- **ICFHR 2018 Competition On Document Image Analysis Tasks for Southeast Asian Palm Leaf Manuscripts**

Publicly available for scientific use on:

http://amadi.univ-lr.fr/ICFHR2016_Contest/

http://amadi.univ-lr.fr/ICDAR2017_Competition/

http://amadi.univ-lr.fr/ICFHR2018_Contest/

Problem Identification on Balinese Palm Leaf Manuscript Images

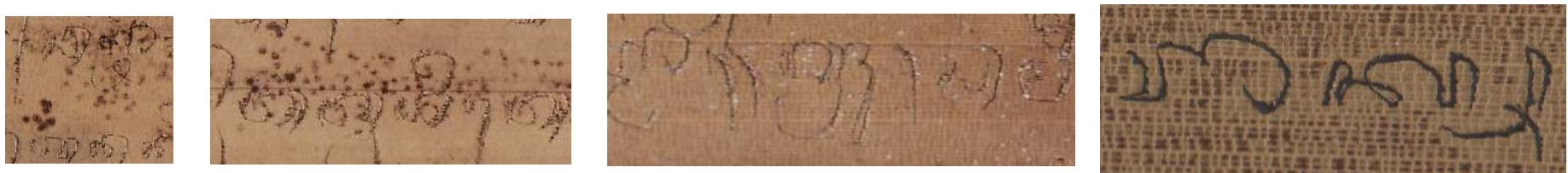
- artifacts due to aging, foxing, yellowing, marks of strain, local shading effects



- low intensity variations, low/poor contrast



- random noises, nonstationary and correlated noises

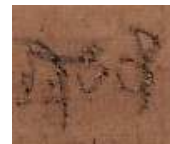


Problem Identification on Balinese Palm Leaf Manuscript Images (2)

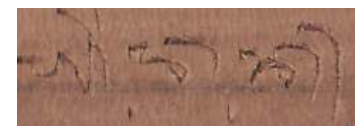
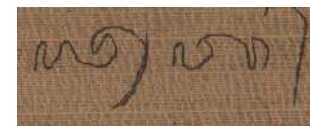
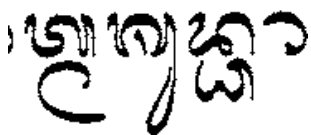
- discoloured document, fading, variance of gray levels within object and the background



- the merges, fractures and other deformations in the character shapes



- use of non standard fonts



Problem Identification on Balinese Palm Leaf Manuscript Images (3)

- varying kerning (space between letters)



- varying leading (space between lines)



- Scanning errors and resolution, problems due to the conversion to digital image



Challenges :: *Technical Tasks*

Technical Challenges

Physical condition of the palm leaf



Low intensity variations, poor contrast



Discoloured document, fading gray levels



Artifacts due to aging, black nuances, foxing, and yellowing



Random noises

Challenging Tasks:

binarization

text line segmentation

glyph segmentation

Challenges :: *Binarization*

Binarization

- hard to separate the text from the background
- extract unrecognizable characters with noise



Method of Otsu, Niblack, Sauvola, Wolf, Rais

Challenges :: *Textline/Glyph Segmentation*

Text Line and Glyph Segmentation



Varying space between lines (leading)



Merges, fractures and other deformations in the character shapes



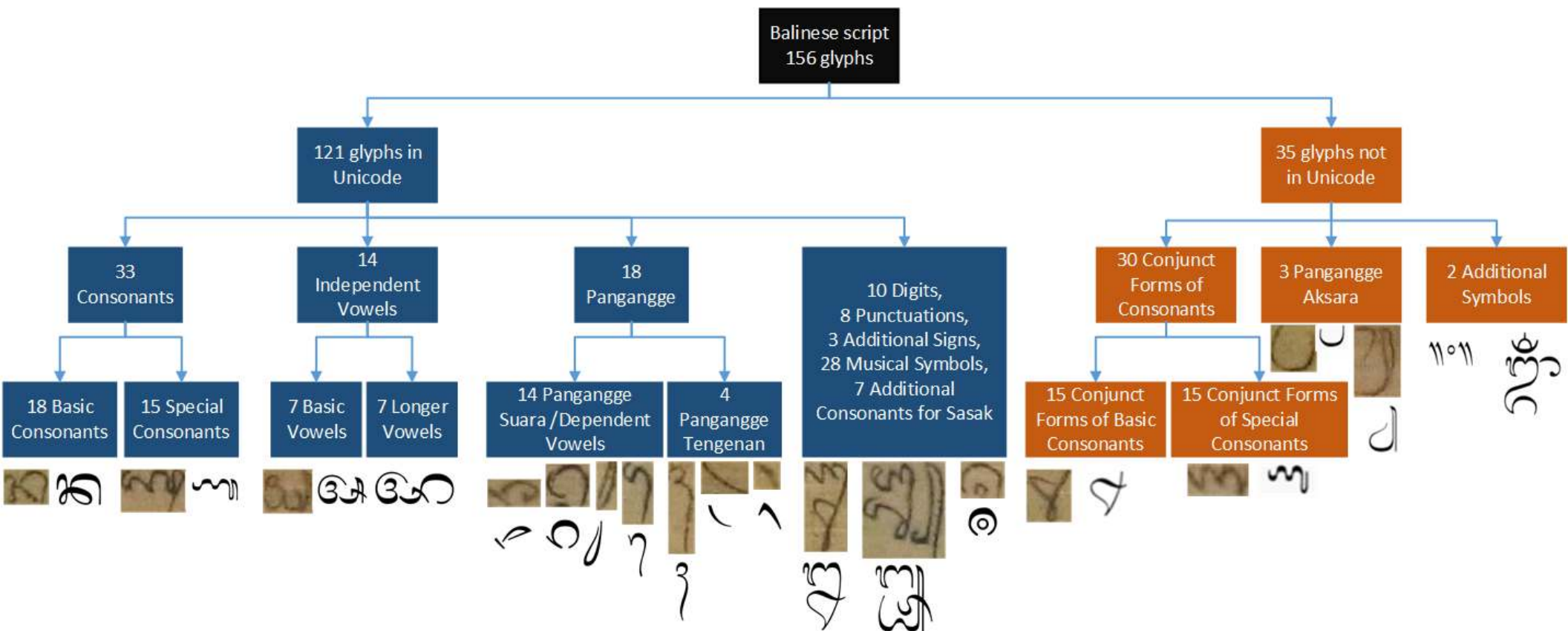
Varying space between glyphs (kerning)

Challenges :: *Textline/Glyph Segmentation*



ပုဂ္ဂိုလ်တို့သည် ဘုရားကိုးကွယ်ရာတွင် နတ်တို့၏ နိဗ္ဗာန်သို့ ရောက်ရှိရန် အားထုတ်ကြသည်။
အားထုတ်ရာတွင် နတ်တို့၏ နိဗ္ဗာန်သို့ ရောက်ရှိရန် အားထုတ်ကြသည်။
အားထုတ်ရာတွင် နတ်တို့၏ နိဗ္ဗာန်သို့ ရောက်ရှိရန် အားထုတ်ကြသည်။
အားထုတ်ရာတွင် နတ်တို့၏ နိဗ္ဗာန်သို့ ရောက်ရှိရန် အားထုတ်ကြသည်။

Challenges :: *Complexity of Balinese Script*



Challenges :: *Complexity of Balinese Script*

wa + na + da = wanada

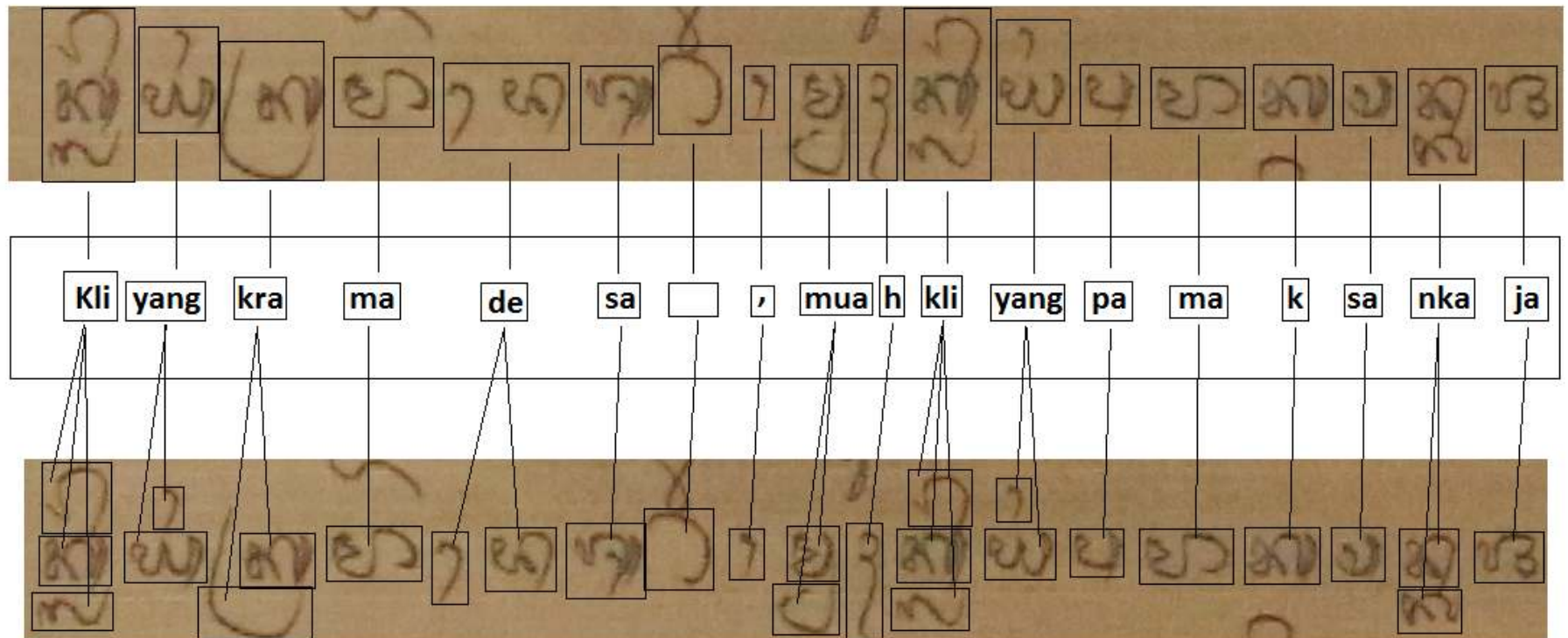
ulu + wa + na + da + suku = wi + na + du = winadu

wi + na + du = wi + ~~na~~ + du = wi + ndu = windu

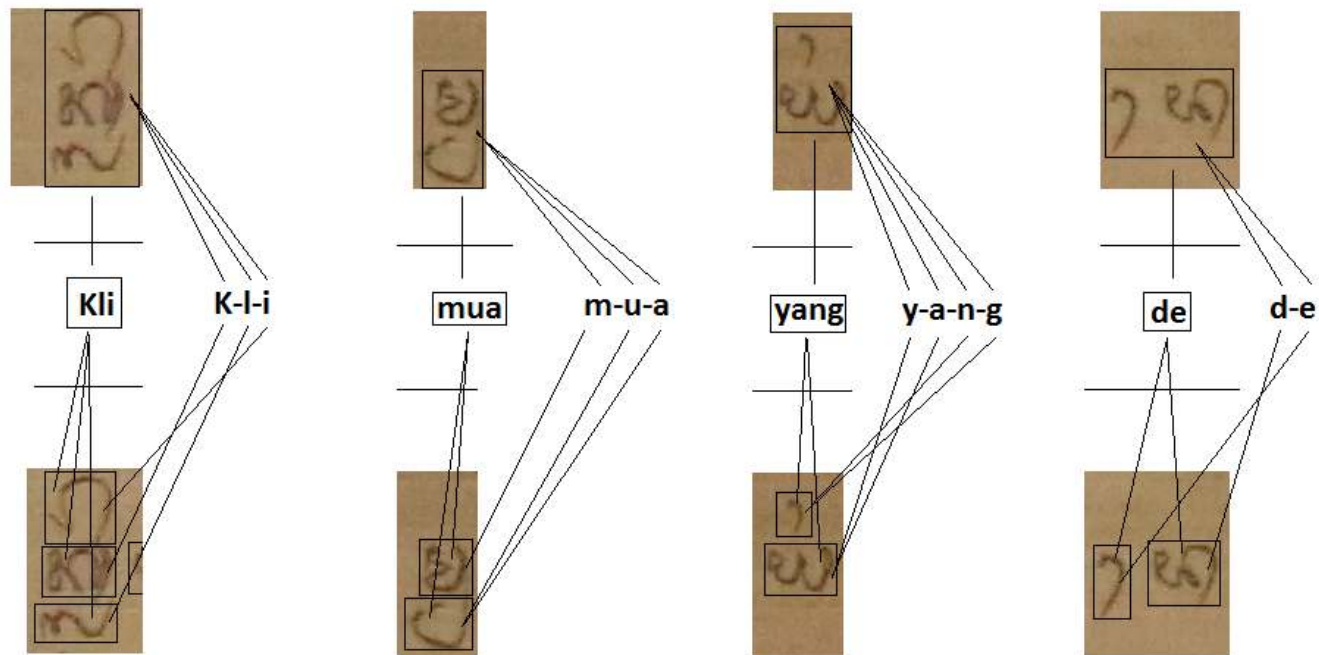
conjunct form of "da"

Challenges :: *Alignment of Balinese Script*

Kliyang krama desa, muah kliyang pamaksan kaja



Challenges :: *Alignment of Balinese Script*



For Balinese script transliteration, there are many :

- Vertical Glyph Arrangement vs Horizontal Character alignments
- Alignment many-to-many

Challenges :: *Character/Glyph Recognition*

Technical Challenges: Complexity of the Balinese script

Challenges in Isolated Glyph Recognition



Different styles in writing Balinese script from different writers



Different proportion size of glyphs



Interclass similarity between glyph



Challenges :: *Character/Glyph Recognition*

'Na'



Model Classifier

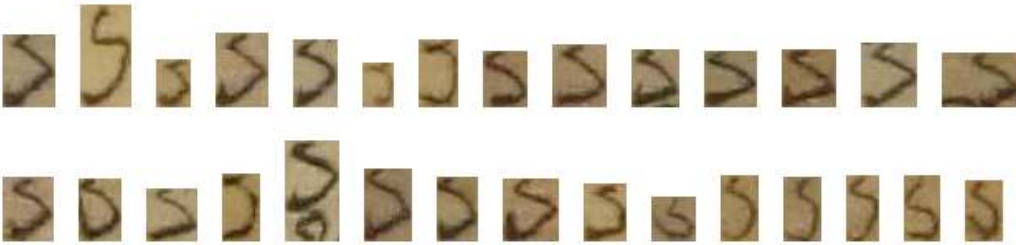
'Ka'



Visual analysis of character appearance

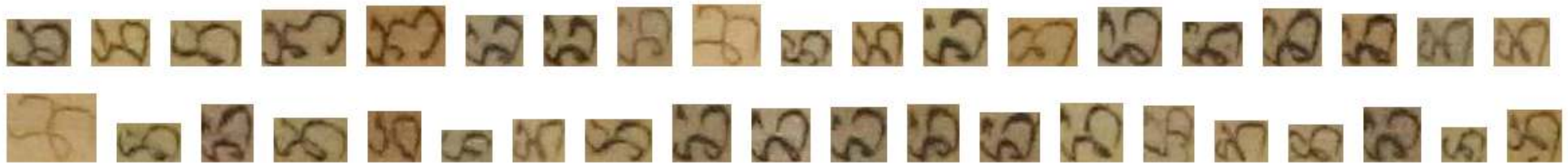
<p>U KARA</p> <p>Download Image(s)</p>	<p>118</p>	
--	------------	---

Visual analysis of character appearance

<p>SUKU ILUT</p> <p>Download Image(s)</p>	55	
---	----	--

Visual analysis of correlation

‘Ba Kembang’



‘Da’



Visual analysis of correlation

‘A Kara’

ଆ



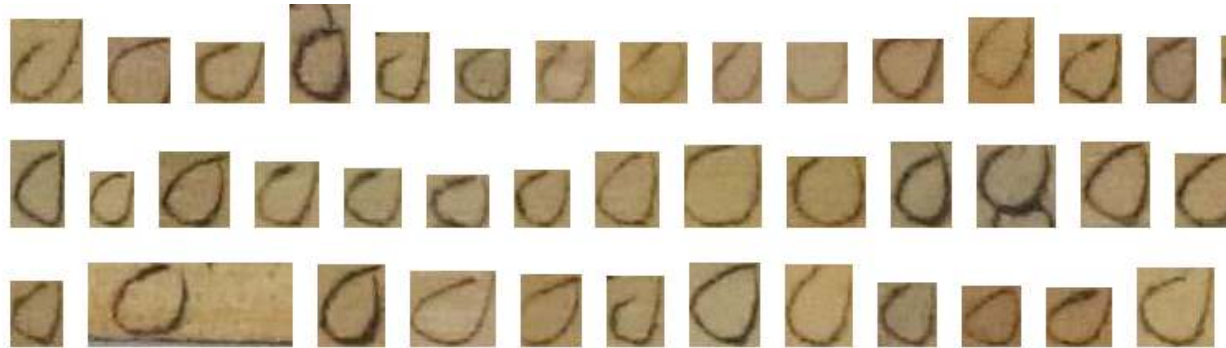
‘Sa’

ସା



Visual analysis of correlation

‘Suku Kembang’



‘Nania’



Visual analysis of correlation

‘Na’



‘Ka’



Visual analysis of correlation

Character	NbTrain	Character	NbTrain	Correlation
BA KEMBANG	31	DA	605	37,5
A KARA	31	SA	294	35,7143
BA	337	A	756	27,7778
NGA	199	DA	605	27,7778
DA	605	NA	1029	25
PEPET	154	A	756	23,5294
NA RAMBAT	88	A	756	22,2222
GA	262	BA	337	21,0526

- Ba Kembang – Da



- Da – Na



- A Kara – Sa



- Pepet - A



- Ba – A



- Na Rambat – A



- Nga – Da



- Ga – Ba



Statistical correlation between characters

Correlation $\geq 20\%$:

Character	NbTrain	Character	NbTrain	Correlation
BA KEMBANG	31	NA	1029	55
SA SAPA	17	SA	294	40
A KARA	31	SA	294	31,5789
SUKU ILUT	35	TEDONG	212	27,7778
DHA MADU	85	CA	115	26,3158
DHA MADU	85	DA	605	21,0526
SA SAGA	51	GA	262	21,0526
NYA	23	BA	337	20



Size of 'Adeg-Adeg' ?



Challenges :: *Word/Text Recognition/Transliteration*

Challenges in Text Transliteration

- syllabic script
- speech sound = certain phonological rules [Antara Kesiman et al., 2018]
- problem of one-to-one mapping between linguistic symbols and images of symbols




Huge combination of possible compound syllable

Problem of "allographs"



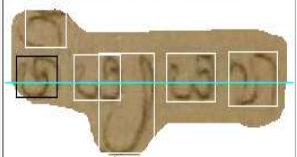
- more than one shape of glyph (image of symbol) is allowed to be used to represent a same sound of speech of syllable (linguistic symbol)[Doermann and Tombre, 2014]

Challenges :: *Word/Text Recognition/Transliteration*



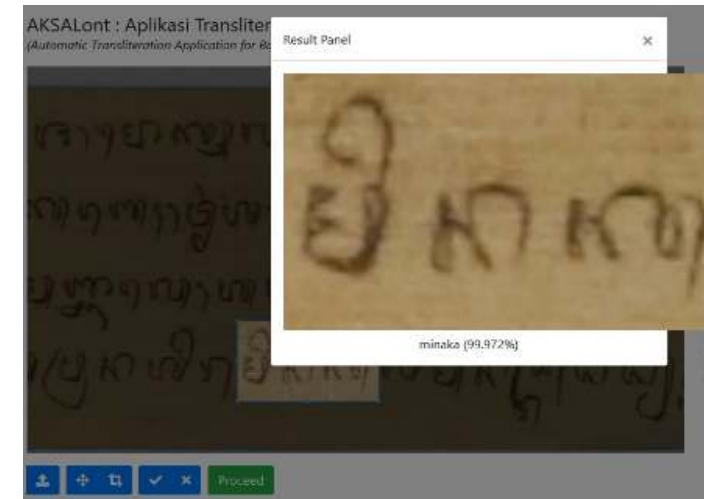
Select ROI to be transliterated:

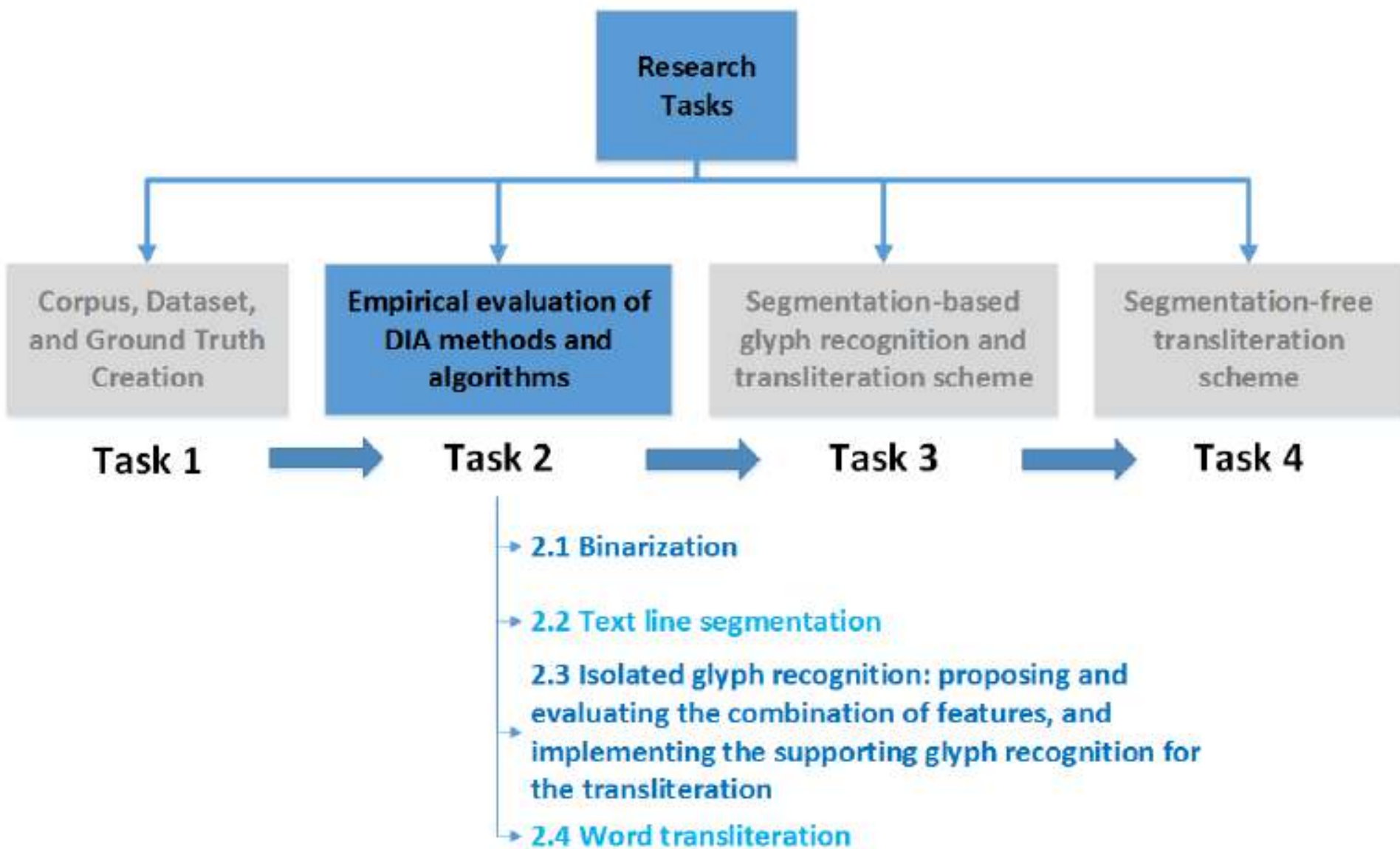
Glyph Segmentation



Glyph Recognition	Transliteration
CC_1_[CECEK GLOBAL 16.3038][CECEK ASC 13.0894][CECEK ASC 13.0894]	
CC_1_CECEK	
CC_2_[GANTUNGAN TA GLOBAL 16.5601][GANTUNGAN TA DESC 21.114][WA BASE 14.4341]	
CC_2_GANTUNGAN TA	
CC_3_[DA MADU GLOBAL 14.2716][DA MADU BASE 10.0468][DA MADU BASE 10.0468]	
CC_3_DA MADU	
CC_4_[NANIA GLOBAL 18.9502][NANIA BASE-DESC 13.1637][NANIA BASE-DESC 13.1637]	
CC_4_NANIA	
CC_5_[DA MADU GLOBAL 15.5316][DA MADU BASE 11.999][DA MADU BASE 11.999]	
CC_5_DA MADU	

DHla*DHara





2.1 Binarization Methods

- Otsu
- Sauvola
- Niblack
- NICK
- Rais
- Wolf
- Howe
- ICFHR G1-G4

2.2 Text Line Segmentation Methods

- Adaptive Partial Projection (APP)
- A* Path Planning
- Shredding method
- Adaptive Local Connectivity Map (ALCM)
- Seam Carving
- Adaptive Path Finding Method

2.3 Isolated Glyph Recognition Methods

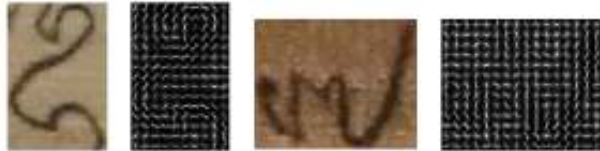
- Handcrafted Feature Extraction Methods:
 - Projection histogram
 - Celled projection
 - Distance profile
 - Crossing
 - Zoning
 - Moments
 - Histogram of Gradient (HoG)
 - Kirsch Directional Edges
 - Neighborhood Pixels Weights (NPW)
- Convolutional Neural Network (CNN)
- Unsupervised Feature Learning (UFL)

2.4 Text Transliteration Methods

Long Short Term Memory Network (LSTM)

Isolated Glyph Recognition (Task 2.3)

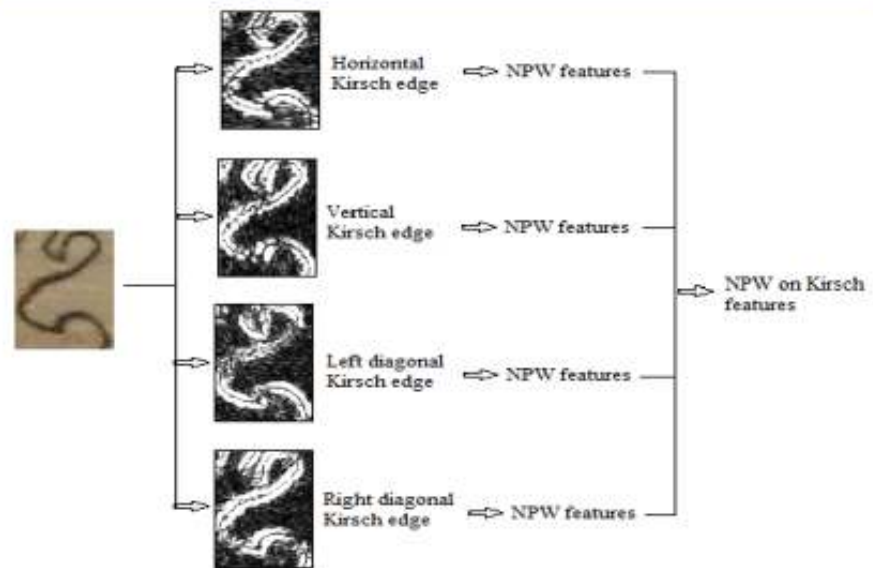
Methods



- **HoG features**
[Aggarwal et al., 2015, Fujisawa et al., 1999], **NPW** [Kumar, 2009], **Kirsch Directional Edges** [Kumar, 2009] and **Zoning** [R.N and Rao, 2014, Blumenstein et al., 2003, Kumar, 2009, Bokser, 1992] separately provide a very promising and good enough result.
- give the initial directional curve features for each glyph
- already serve as a good feature discriminants for Balinese script glyphs

Proposed Combination of Features

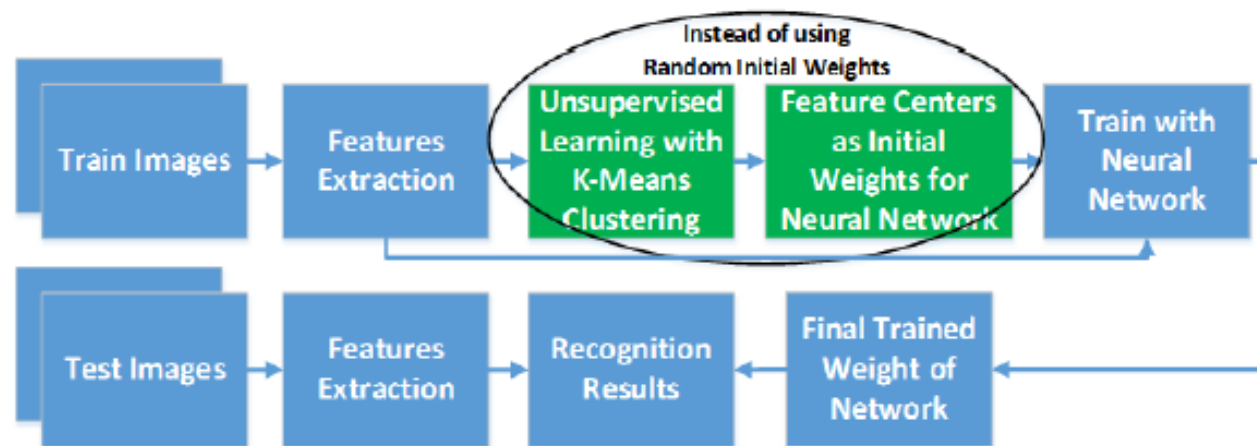
[Kesiman et al., 2016b]



NPW on Kirsch features + HoG +
Zoning

Isolated Glyph Recognition (Task 2.3)

Schema of glyph recognizer with feature extraction method, unsupervised feature learning and neural network [Kesiman et al., 2017]



- inspired by [Coates et al., , Coates et al., 2011].
- sent combination of feature vector to a single layer neural network
- applied also an additional sub module for the initial unsupervised learning based on K-Means clustering
- The unsupervised learning calculates the initial learning weight for the neural network training phase from the cluster centres of all feature vectors

Isolated Glyph Recognition (Task 2.3)

Evaluation Metrics

the recognition rate = \mathbf{C}/\mathbf{N} , where \mathbf{C} is the number of correctly recognized samples, and \mathbf{N} is the total number of test samples [Burie et al., 2016].

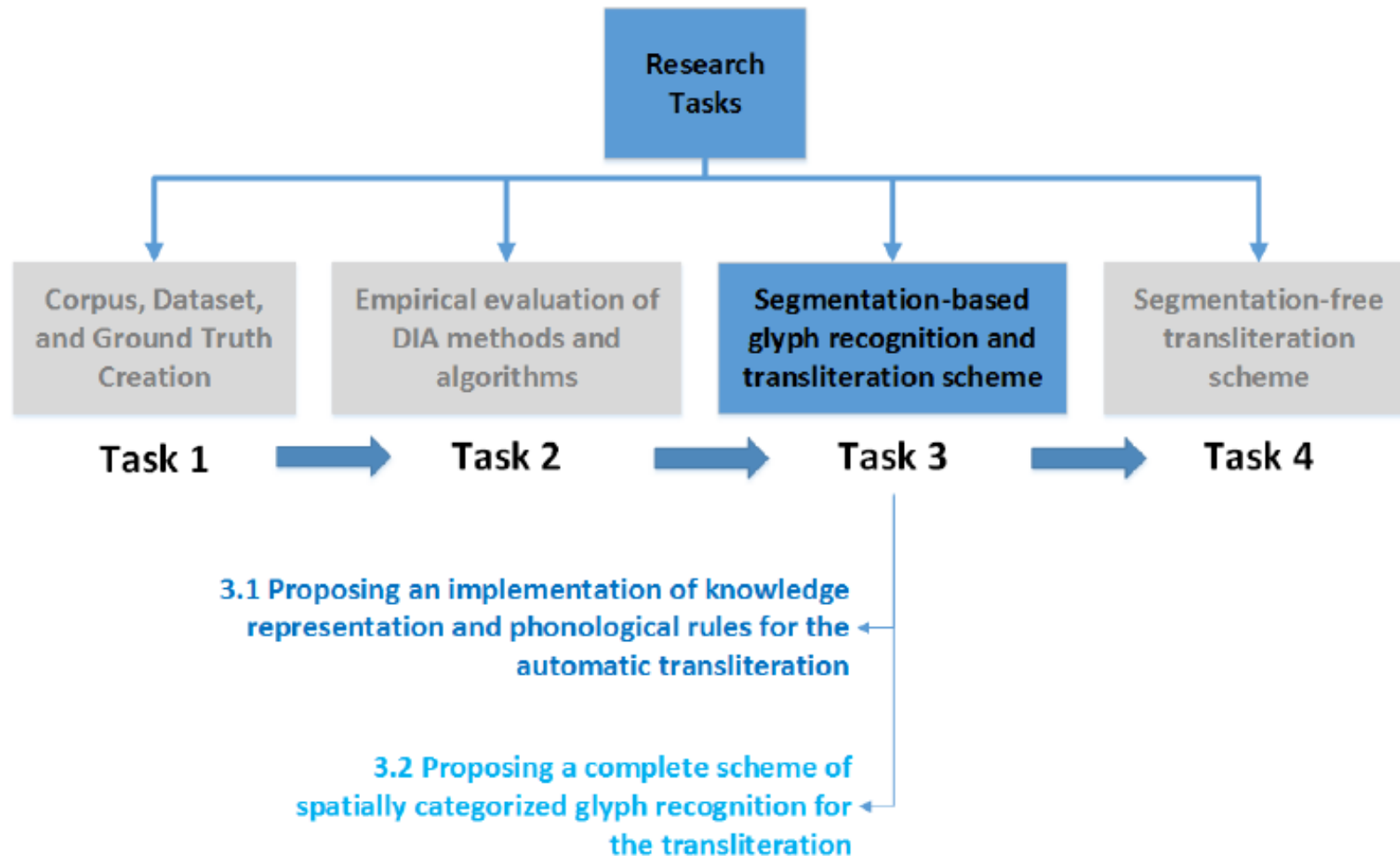
Recognition rate of the global glyph recognizer

Glyph Classes	Number of Data	CNN 1	k-NN	NN	NN UFL
133	Train : 11,710, Test : 7,673	84.31	85.16	85.51	85.63

Recognition rate of the categorized glyph recognizer

Category	Glyph Classes	Number of Data Subsets	NN	NN UFL
ASC	7	Train : 860, Test : 921	92.73	93.16
DESC	20	Train : 1,860, Test : 593	85.84	88.03
BASE	49	Train : 5,070, Test : 4,392	87.46	87.43
ASC-BASE	16	Train : 1,170, Test : 208	75.48	75.96
BASE-DESC	40	Train : 2,550, Test : 1,309	86.40	86.63

Segmentation-based Glyph Recognition and Transliteration Scheme



Knowledge Representation (Task 3.1)

Formalizing syllables: Speech sound = Onset+Nucleus+Coda

No	Consonant basic glyph name	Second glyph form name	Speech sound	Onset	Nucleus
1.	NA and NA TEDONG	GANTUNGAN NA	NA	N	A
4.	A and A TEDONG	GANTUNGAN A	A		
31.	U KARA		U		



Consonant basic glyphs and their second glyph form (conjunct form)

No	Consonant compound glyph name	Glyph component	Speech sound	Onset	Nucleus	Coda
1.	TU	TA + SUKU	TU	T	U	-
3.	I		I			-
4.	NI	NA + ULU	NI	N	I	-
13.	NING	NA + ULU + CECEK	NING	N	I	NG
46.	GRA	GA + GUWUNG	GRA	GR	A	-



Consonant compound glyphs

No	Glyph name	Type	Speech sound
1	0 - 9	Numeral	0-9
2	PAMADA	Punctuation	. (point)
3	TALING	Vowel	E
9	SURANG	Special Consonant	R
10	ADEG-ADEG	Special Consonant	-
16	SUKU ILUT	Vowel	U



Numeral, Punctuation, Vowel, and Special Consonant Glyphs

Alphabet / Dictionary for Glyph Category

```
<?xml version="1.0" encoding="ISO-8859-1"?>
```

```
<BIBLIOGRAPHY>
```

```
<OBJECT>
```

```
<CLASS>
```

```
<LEVEL>Level1</LEVEL> <NAME>TALENG</NAME>
```

```
<LEVEL>Level2</LEVEL> <NAME>VOC</NAME>
```

```
<LEVEL>Level3</LEVEL> <NAME>BASE-DESC</NAME>
```

```
</CLASS>
```

```
<PARAMETER>id</PARAMETER> <VALUE>1</VALUE>
```

```
<PARAMETER>sound</PARAMETER> <VALUE>E</VALUE>
```

```
<PARAMETER>end</PARAMETER> <VALUE>*</VALUE>
```

```
<PARAMETER>split</PARAMETER> <VALUE>*</VALUE>
```

```
</OBJECT>
```

```
<OBJECT>
```

```
<CLASS>
```

```
<LEVEL>Level1</LEVEL> <NAME>NA</NAME>
```

```
<LEVEL>Level2</LEVEL> <NAME>CON</NAME>
```

```
<LEVEL>Level3</LEVEL> <NAME>BASE</NAME>
```

```
</CLASS>
```

```
<PARAMETER>id</PARAMETER> <VALUE>2</VALUE>
```

```
<PARAMETER>sound</PARAMETER> <VALUE>N</VALUE>
```

```
<PARAMETER>end</PARAMETER> <VALUE>A</VALUE>
```

```
<PARAMETER>split</PARAMETER> <VALUE>*</VALUE>
```

```
</OBJECT>
```

```
.....
```

```
</BIBLIOGRAPHY>
```

Level1 : Name of the glyph

Level2 : VOC, CON, GEM, GAN, NUM, PUN

Level3 : SPATIAL Information of the glyph

(ASC, ASC-BASE, BASE, BASE-DESC, DESC, ASC-BASE-DESC)

Id : class number of the glyph

Sound : root sound of the glyph

End : end sound of the glyph

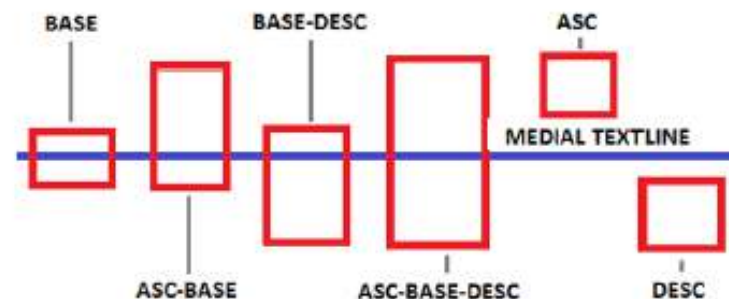
Split : compound sound of the compound glyph

Level1	Level2	Level3	Id	Sound	End	Split
TALENG	VOC	BASE-DESC	1	E	*	*
NA	CON	BASE	2	N	A	*
KA	CON	BASE	3	K	A	*
TA	CON	BASE	4	T	A	*
A	CON	BASE	5		A	*
ULU	VOC	ASC	6	I	*	*
CECEK	VOC	ASC	7	NG	*	*
WA	CON	BASE	8	W	A	*
DA	CON	BASE	9	D	A	*
ADEG-ADEG	GEM	ASC-BASE-DESC	10	*	*	*
JA	CON	BASE	11	J	A	*
BISAH	VOC	BASE-DESC	12	H	*	*
LA	CON	BASE	13	L	A	*

Knowledge Representation (Task 3.1)

Building Glyph Dictionary: Glyph Properties and Categorizations

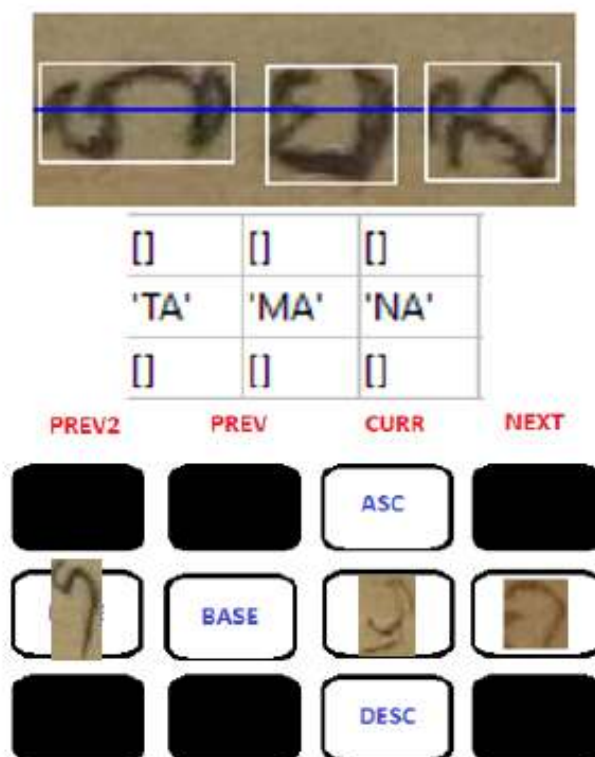
- Property "Id": identity number of the glyphs.
- Property "Level1": name of the glyphs.
- Property "Level2" is categorized in six groups:
 - CON for consonant,
 - VOC for vocal,
 - GAN for gantungan (conjunct form),
 - GEM for gempelan (conjunct form),
 - NUM for numeral, and
 - PUN for punctuation.
- Property "Level3": spatial position of the glyphs
- Property "StartSyllable": onset of syllable for the consonant basic glyphs or the speech sound for the consonant compound glyphs, numeral, punctuation, and special consonant glyphs.
- Property "EndSyllable": nucleus of syllable for the consonant basic glyphs.
- Property "SplitSyllable": onset, nucleus, and coda of syllable for the consonant compound glyphs.



Phonological Rules (34 Rules) (Task 3.1)

Contribution: Formalization of phonological rules [Antara Kesiman et al., 2018]

- based on phonetics of traditional linguistic study of Balinese transliteration
- applied in sequential conditional checking order
- the final speech sound for a syllable of a current (CURR) base (BASE) glyph will be determined by :
 - the ascender (**ASC**) of current glyph,
 - the descender (**DESC**) of current glyph,
 - the **BASE** of the **NEXT** glyph,
 - the **BASE** of the previous (**PREV**) glyph,
 - the **BASE** of the two previous (**PREV2**) glyphs.



Example: a rule for TALING and TEDONG

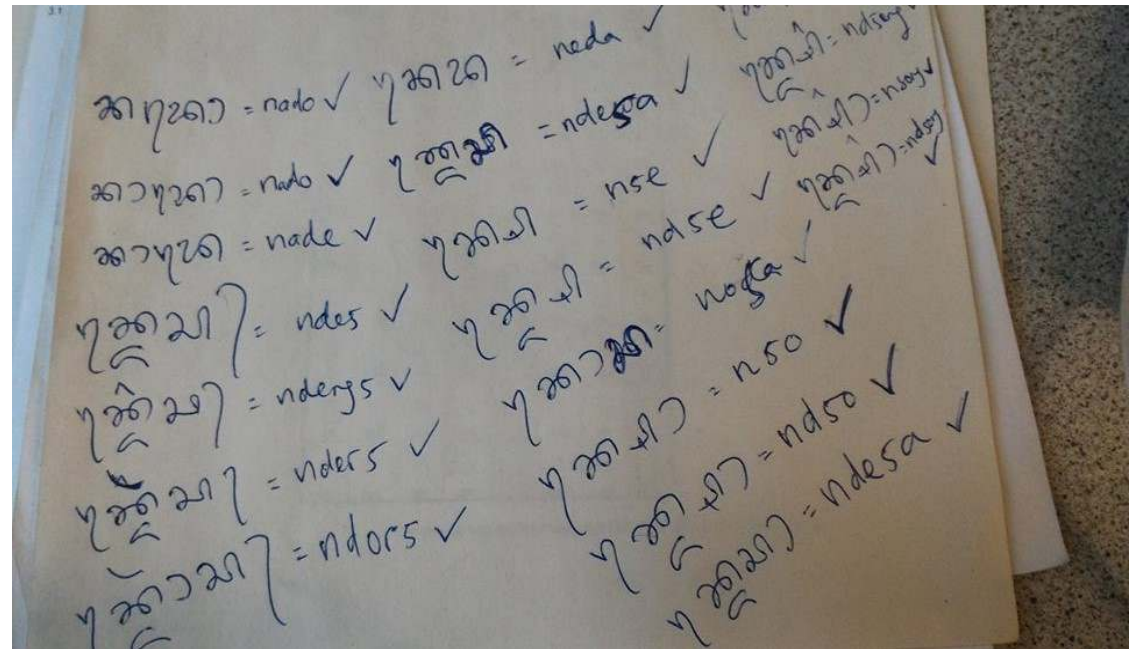
RULE6: IF PREV2.BASE.LEVEL1 = "TALING" AND CURR.BASE.LEVEL1 ≠ EMPTY AND CURR.BASE.LEVEL2 ≠ CON AND CURR.BASE.LEVEL2 = GEM AND CURR.BASE.LEVEL3 = BASE AND NEXT.BASE.LEVEL1 = "TEDONG" ⇒ SPEECH_SOUND = SPEECH_SOUND + "O"

Grammatical Rules

- Example :

- “The CONSONANT can apply the rule of TALENG and/or TEDONG if his next is not a GEMPELAN. If his next is a GEMPELAN, than the rule of TALENG and/or TEDONG will be applied for that next GEMPELAN. Meanwhile, the GEMPELAN can take into account the rule of TALENG and/or TEDONG if and only if TALENG can be found in the two previous position of this GEMPELAN.”

Defined 34 grammatical rules

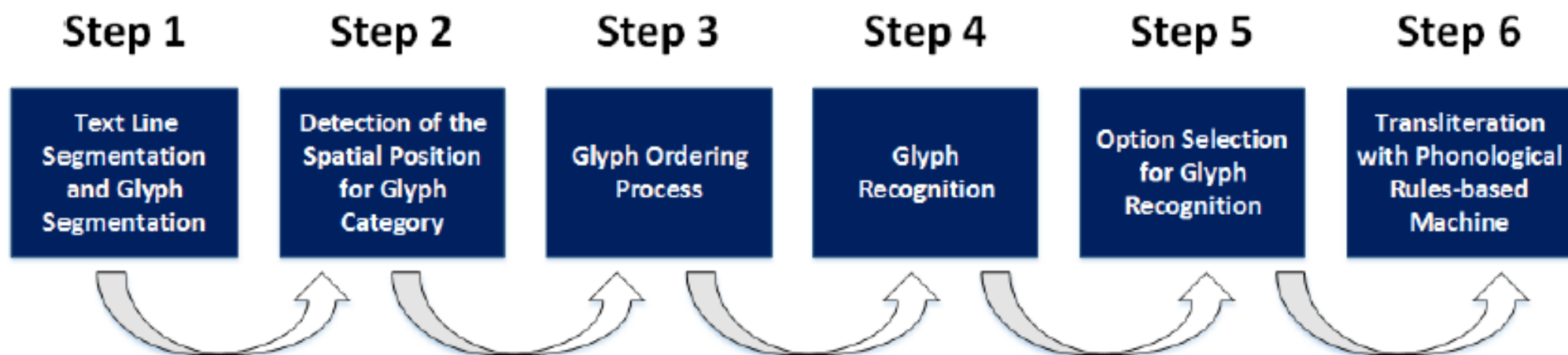


Grammatical Rules

NO	PREV2	PREV	CURR								NEXT		RESULT
	BASE	BASE	ASC	BASE					DESC		BASE		
	LEVEL1	LEVEL1	LEVEL1	LEVEL1	LEVEL2	LEVEL3	END	SPLIT	LEVEL1	SOUND	LEVEL1	LEVEL2	
1				~EMPTY	CON/GEM	BASE							result=CURR_SOUND_BASE
2				~EMPTY	CON/GEM	BASE			~EMPTY				result=result+CURR_SOUND_DESC
3		TALENG		~EMPTY	CON&~GEM	BASE					~TEDONG		result=result+PREV_SOUND_BASE
4		TALENG		~EMPTY	CON&~GEM	BASE					TEDONG		result=result+"o"
5	TALENG			~EMPTY	~CON&GEM	BASE					~TEDONG		result=result+"e"
6	TALENG			~EMPTY	~CON&GEM	BASE					TEDONG		result=result+"o"
7				~EMPTY	CON/GEM	BASE					NANIA		result=result+NEXT_SOUND_BASE
8		~TALENG	EMPTY	~EMPTY	CON	BASE	"A"		EMPTY		~ADEG-ADEG	~GEM	result=result+"a"
9		~TALENG	EMPTY	~EMPTY	GEM	BASE	"A"		EMPTY		~ADEG-ADEG	~GEM	result=result+"a"
10		~TALENG	CECEK/SURANG	~EMPTY	CON	BASE	"A"		EMPTY		~ADEG-ADEG	~GEM	result=result+"a"
11		~TALENG	CECEK/SURANG	~EMPTY	GEM	BASE	"A"		EMPTY		~ADEG-ADEG	~GEM	result=result+"a"
12		~TALENG	EMPTY	~EMPTY	CON/GEM	BASE			~EMPTY	"A"	~ADEG-ADEG	~GEM	result=result+"a"
13		~TALENG	CECEK/SURANG	~EMPTY	CON/GEM	BASE			~EMPTY	"A"	~ADEG-ADEG	~GEM	result=result+"a"
14			~EMPTY	~EMPTY	CON/GEM	BASE							result=result+CURR_SOUND_ASC
15				~EMPTY	CON/GEM	BASE					BISAH		result=result+NEXT_SOUND_BASE
16				~EMPTY	CON	ASC-BASE/BASE-DESC		~"B", root_part=split_part1(CURR_SPLIT_BASE)	~EMPTY				result=root_part
				~EMPTY	CON	ASC-BASE/BASE-DESC		"*", root_part=""	~EMPTY				result=root_part
17				~EMPTY	CON	ASC-BASE/BASE-DESC			EMPTY				result=CURR_SOUND_BASE
18				~EMPTY	CON	ASC-BASE/BASE-DESC			~EMPTY				result=result+CURR_SOUND_DESC
19		TALENG		~EMPTY	CON	ASC-BASE/BASE-DESC					~TEDONG	~GEM	result=result+PREV_SOUND_BASE
20		TALENG		~EMPTY	CON	ASC-BASE/BASE-DESC					TEDONG	~GEM	result=result+"o"

A Complete Scheme Of Spatially Categorized Glyph Recognition For The Transliteration (Task 3.2)

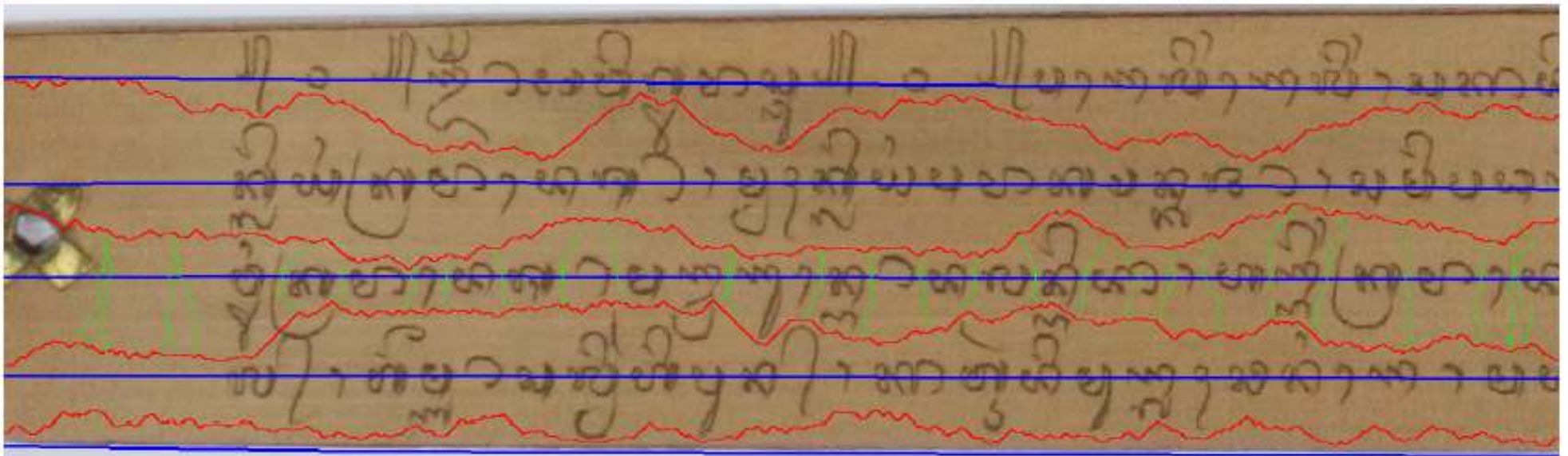
Contribution: Proposed Scheme [Kesiman et al., 2017]



STEP 1: Text Line and Glyph Segmentation (Task 3.2)

Two types of seams are calculated: the medial seams and separating seams.

Seam carving method [Arvanitopoulos and Susstrunk, 2014]

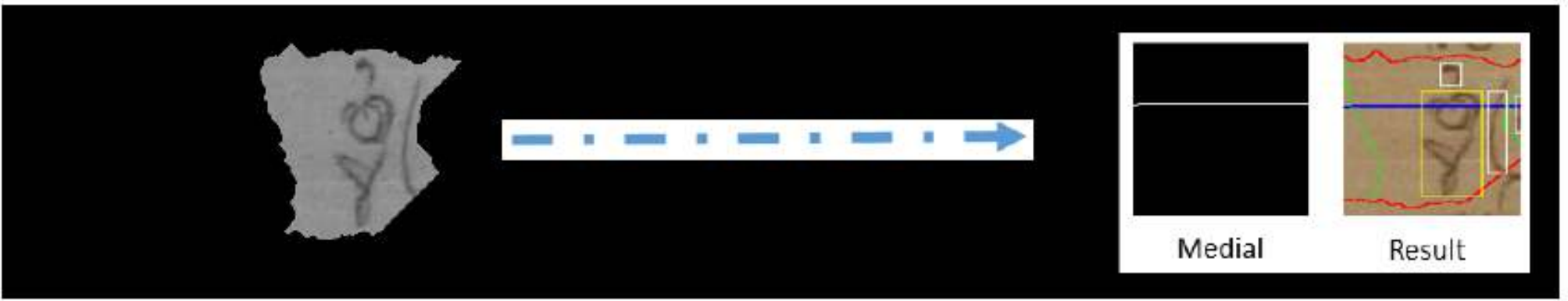
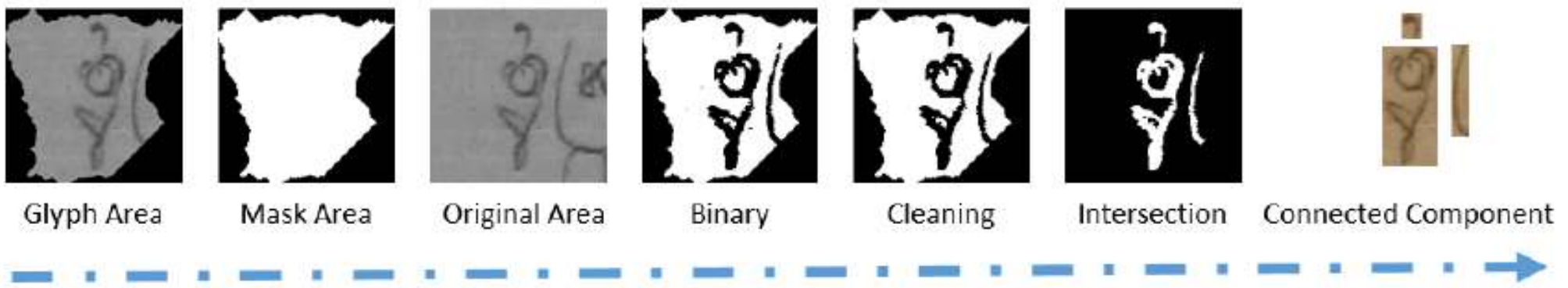


Original code : <https://www.epfl.ch/labs/ivrl/research/handwriting-recognition/text-line-extraction/>

Applied seam carving also for glyph area detection

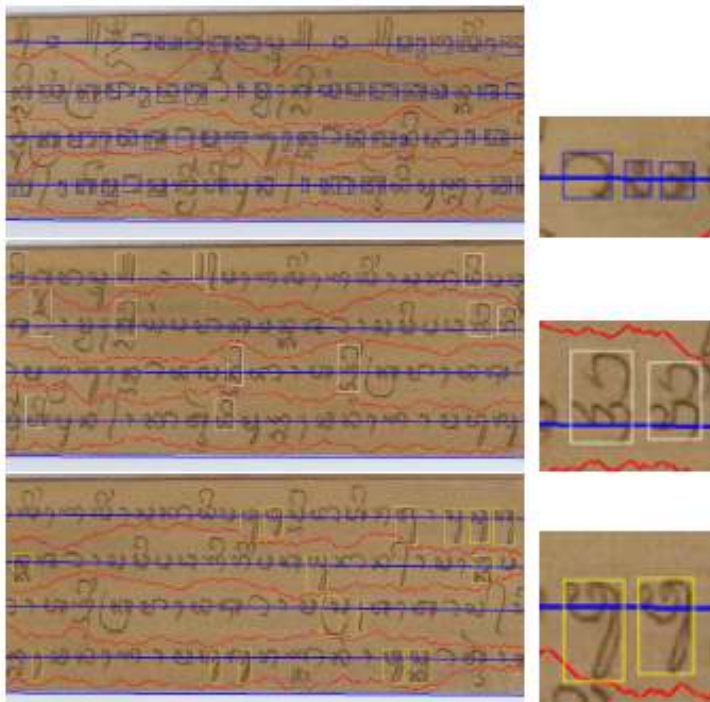
STEP 1: Text Line and Glyph Segmentation (Task 3.2)

Glyph area detection and glyph segmentation process



STEP 2: Detection of the Spatial Position for Glyph Category (Task 3.2)

6 spatial positions for categorized glyph recognizer and phonological rules



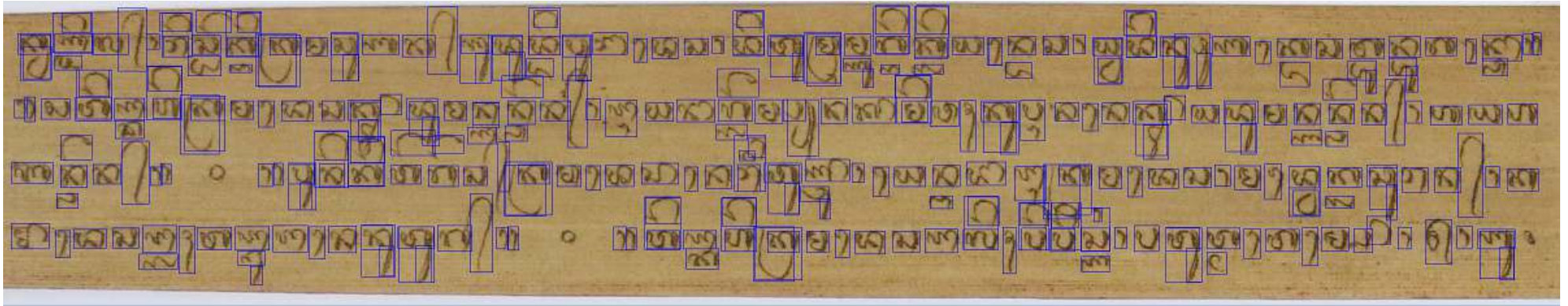
BASE Glyphs, ASC-BASE Glyphs, DESC-BASE Glyphs



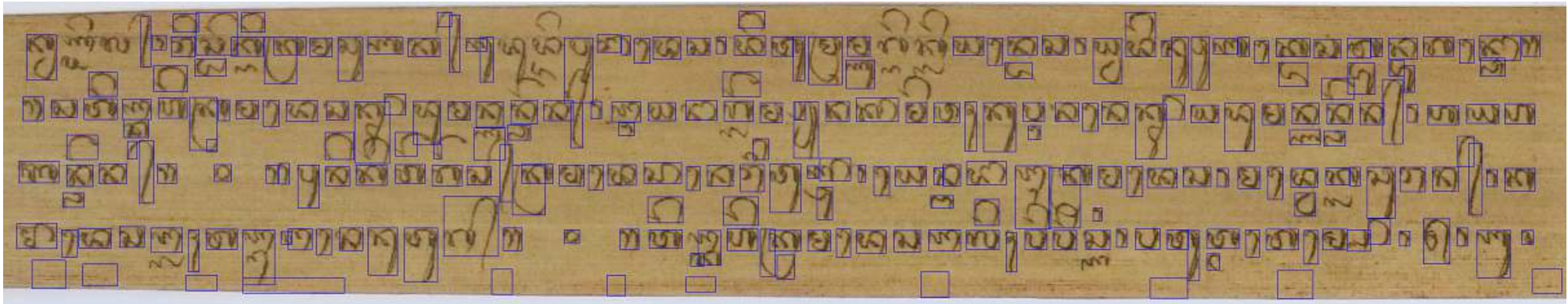
ASC-BASE-DESC Glyphs, ASC Glyphs, DESC Glyphs

Evaluation of Glyph Segmentation and Glyph Recognition

Ground Truth Glyph Segments

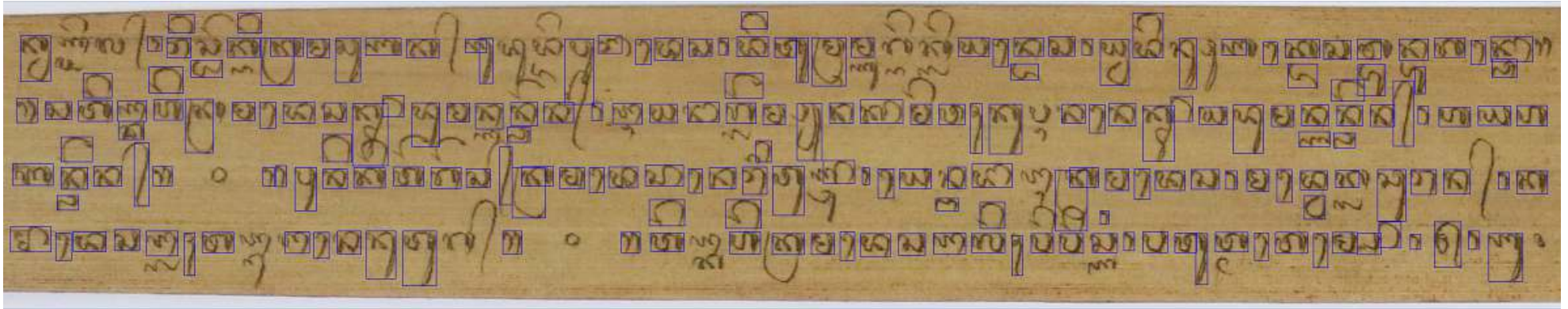


Result of Glyph Segments

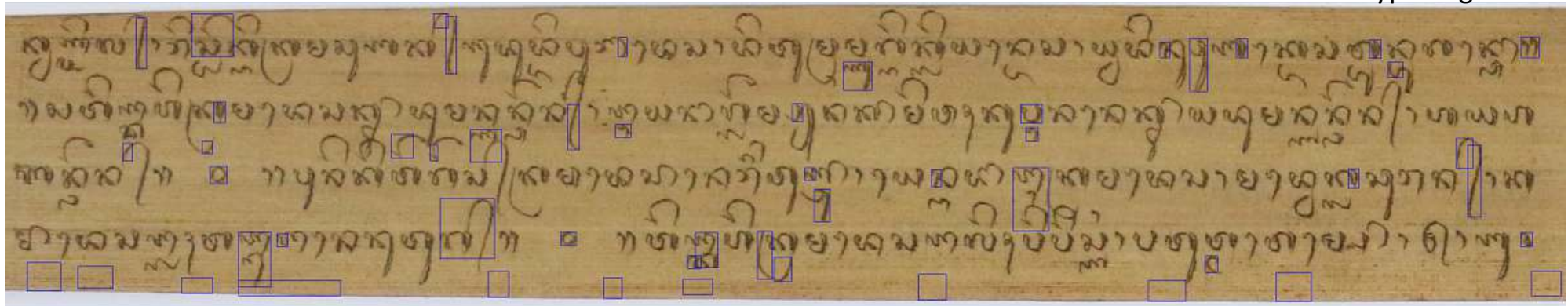


Evaluation of Glyph Segmentation and Glyph Recognition

Correct Overlapped Glyph Segments



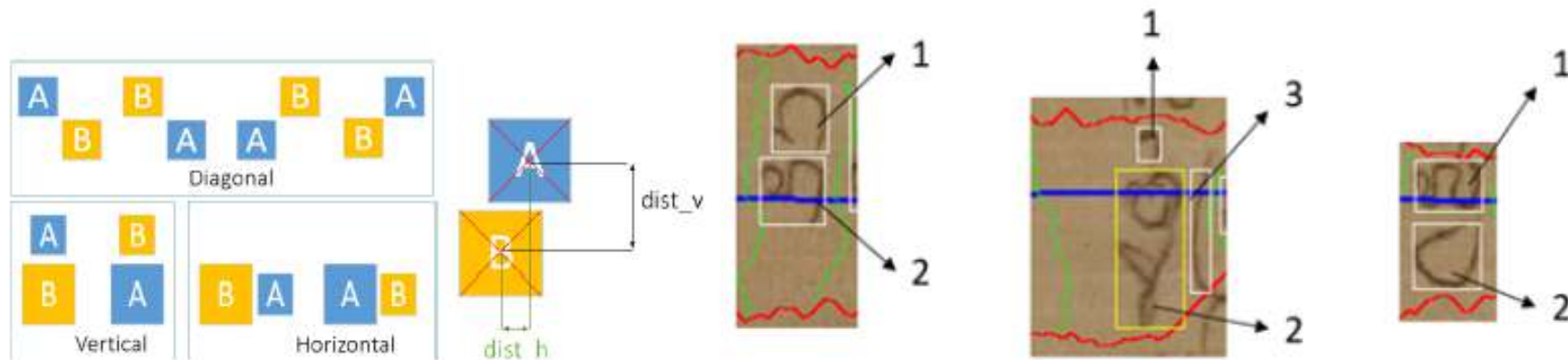
Uncorrect Glyph Segments



STEP 3: Glyph Ordering Process (Task 3.2)

Ordering rule: “BASE-ASC-BASE-DESC-BASE Order”

- glyphs on the medial text line are ordered from left to right based on their left border position on the glyph area
- Spatial Relation between Glyphs:** the **ASC** → before their associate **BASE** glyph, and the **DESC** glyph → after their associate **BASE** glyph



Spatial Relation between Glyphs: In the second example, glyph 1 (**ASC**) and glyph 2 (**BASE-DESC**): vertical relation, glyph 1 and glyph 3 (**BASE-DESC**):diagonal relation, and glyph 2 and glyph 3: horizontal relation. Glyph 1 belongs to glyph 2, glyph 1 does not belong to glyph 3

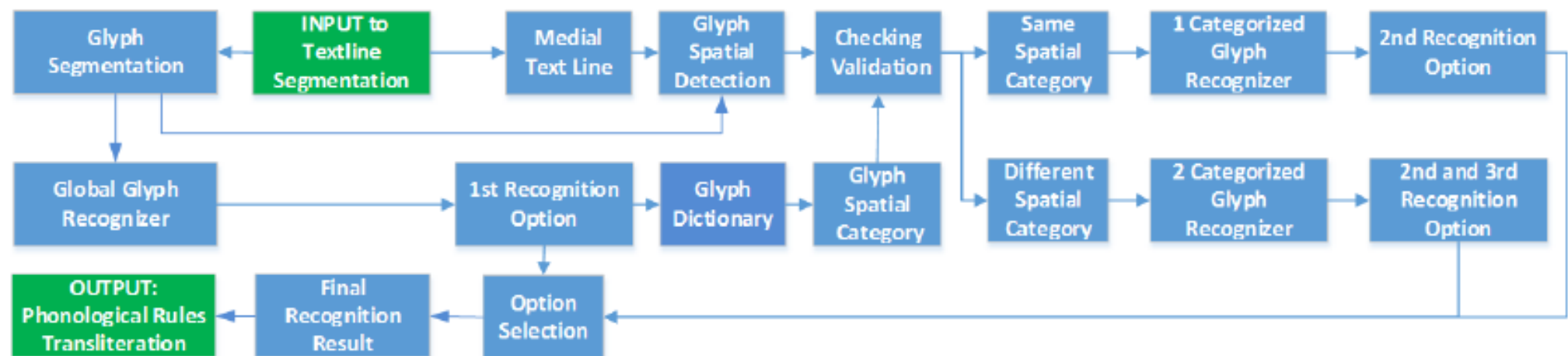
STEP 4: Glyph Recognition (Task 3.2)

Global Glyph Recognition and Categorized Glyph Recognition

- using the same glyph recognizer schema from Task 2.3
- one global glyph recognizer and five different categorized glyph recognizers were built.
 - global recognizer: complete 133 glyph classes
 - five different categorized glyph recognizers: subset of glyph classes for each different spatial position category
 - ① BASE
 - ② ASC
 - ③ DESC
 - ④ ASC-BASE
 - ⑤ BASE-DESC

STEP 5: Option Selection for Glyph Recognition (Task 3.2)

Recognition Options



- Global Recognition (**G**)
- Categorized Recognition based on Glyph Dictionary (**D**)
- Categorized Recognition based on Glyph Spatial Position (**S**)

STEP 5: Option Selection for Glyph Recognition (Task 3.2)

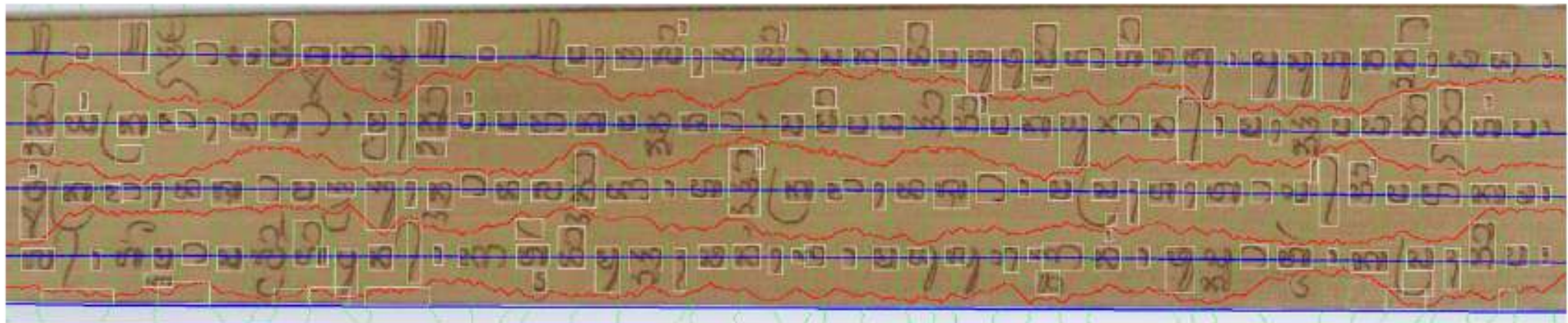
Option selection rules

- If $D = S$, there are two possibilities:
 - If $G = D = S$, the final recognition is $G/D/S$
 - If $G \neq (D = S)$, the final recognition is D/S
- If $D \neq S$, there are three possibilities:
 - If $(G = S) \neq D$, impossible case
 - If $(G = D) \neq S$,
 - If $(S = \text{BASE}/\text{ASC-BASE}/\text{BASE-DESC}/\text{ASC-BASE-DESC})$ and $D = \text{ASC}/\text{DESC})$ or vice-versa, the final recognition is S
 - If $(S = \text{ASC})$ and $D = \text{DESC})$ or vice-versa, the final recognition is G/D
 - For all other sub-cases, it may be a bad glyph segmentation. The final recognition is G/D
- If $G \neq D \neq S$, the final recognition is G

STEP 5: Option Selection for Glyph Recognition (Task 3.2)

Example 1: Option selection rules

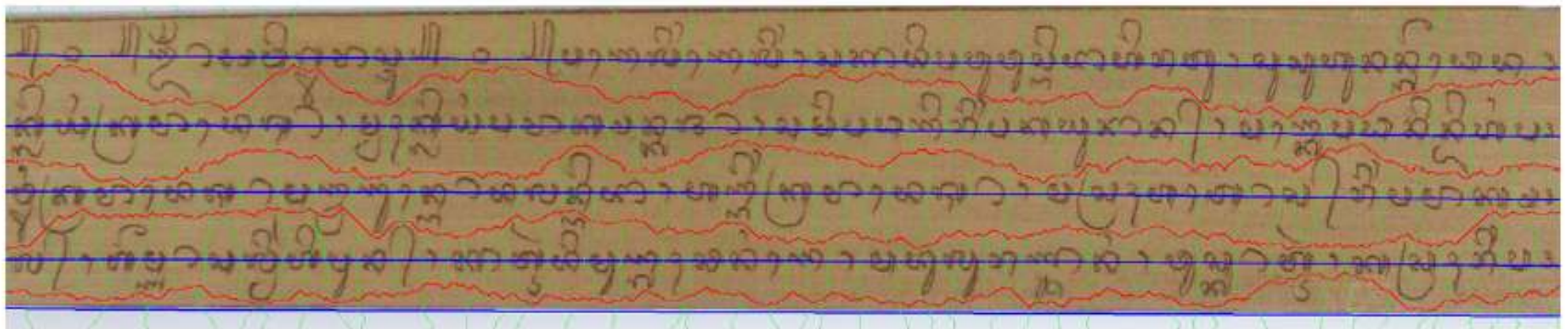
- If the spatial category detection is the same with the glyph dictionary, there are two possibilities:
 - If $\mathbf{G} = \mathbf{D} = \mathbf{S}$, there is only one option. It is a high confidence of correct segmentation and recognition. The final recognition result is $\mathbf{G}/\mathbf{D}/\mathbf{S}$



STEP 5: Option Selection for Glyph Recognition (Task 3.2)

Example 2: Option selection rules

- If the spatial category detection is different with the glyph dictionary, there are three possibilities:
 - If $(\mathbf{G} = \mathbf{D}) \neq \mathbf{S}$, there are two different options. There are three sub-cases.
 - If $(\mathbf{S} = \text{BASE/ASC-BASE/BASE-DESC/ASC-BASE-DESC})$ and $\mathbf{D} = \text{ASC/DESC}$ or vice-versa, it means that there are a big difference between spatial category detection and the glyph dictionary. The final recognition result is \mathbf{S}



Evaluation: Glyph Segmentation and Recognition (Task 3.2)

Evaluation Metrics

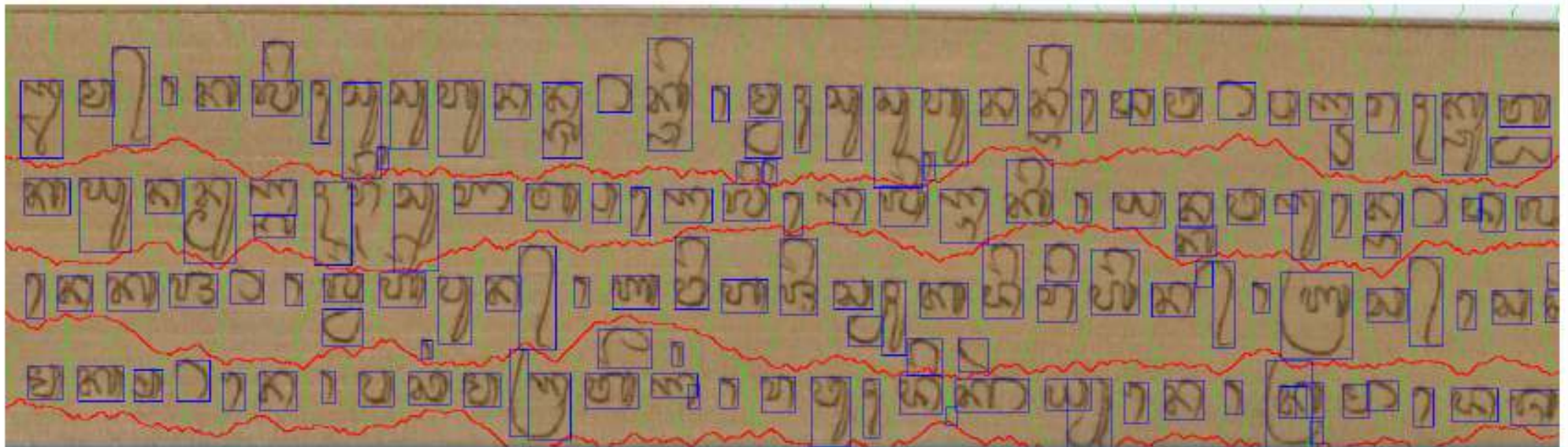
- Number Segments Result (**NSR**) = total number of glyph segments in result file
- Number Segments Overlapped (**NSO**) = number of correctly overlapped (>50%) glyph segments between result file and ground truth file
- Number Recognized Result (**NRR**) = number of correctly recognized glyph segments in result file
- Segmentation Rate (**SR**) = the percentage of **NSO** / **NSR**
- Segmented Recognition Rate (**SRR**) = the percentage of **NRR** / **NSO**



Top left: ground truth, Bottom left: glyph segments result, Top right: correctly overlapped glyph segments, Bottom right: wrong glyph segments

Evaluation: Glyph Segmentation and Recognition (Task 3.2)

Results

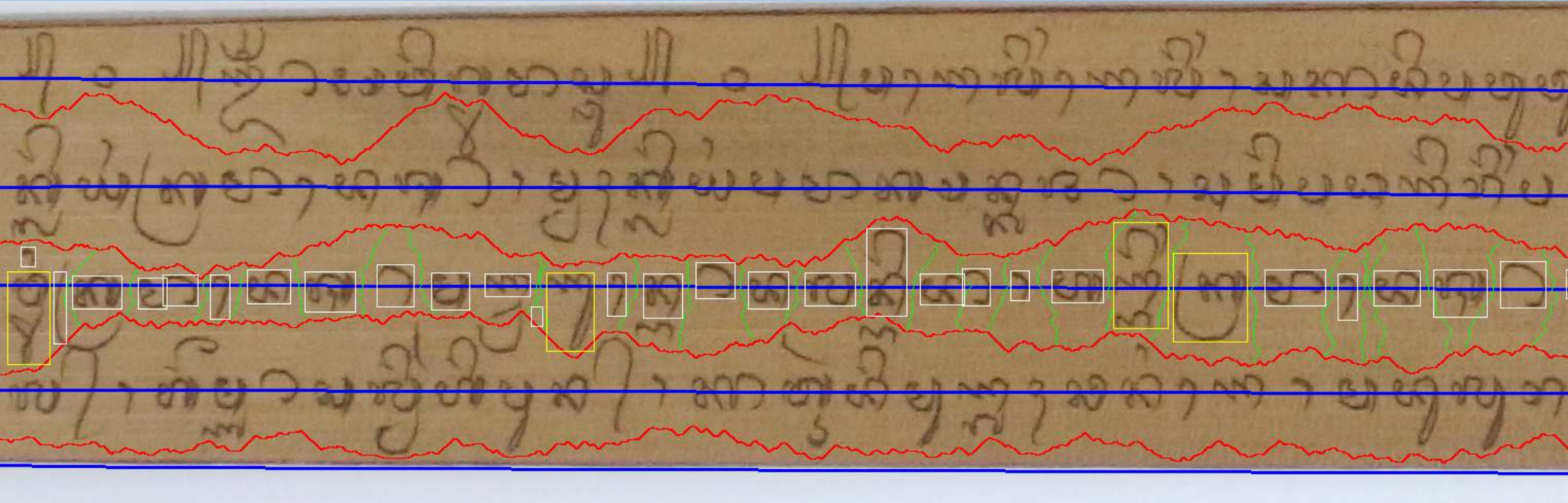


IIA-5-789-P2.jpg.txt - Notepad

File Edit Format View Help

```
[ ] [ ] [2] [MA] [Adeg-Adeg] [CECEK] [KA] [LA] [ULU] [BISAH] [NANIA] [CECEK] [SU] [TU] [RA] [RI] [TEDONG] [RI]
TALENG] [KI] [ ] [PA] [DA MADU] [NI] [GANTUNGAN DA] [NI] [A] [CECEK] [PA] [SUKU ILUT] [ ] [ ] [ ] [KA] [YU] [NA]
A] [MA TEDONG] [KA] [GEMPELAN SA SAPA] [4] [ ] [ ] [ ] [CECEK] [NA] [KA] [JA] [TEDONG] [BISAH] [SUKU KEMBUNG] [LA]
] [TEDONG] [WA] [TA] [SUKU] [BISAH] [KA] [SUKU KEMBUNG] [SUKU] [SA] [BISAH] [RI] [PA] [4] [ ] [ ] [ ] [ ] [ ] [MA] [K
```

Checking Validation

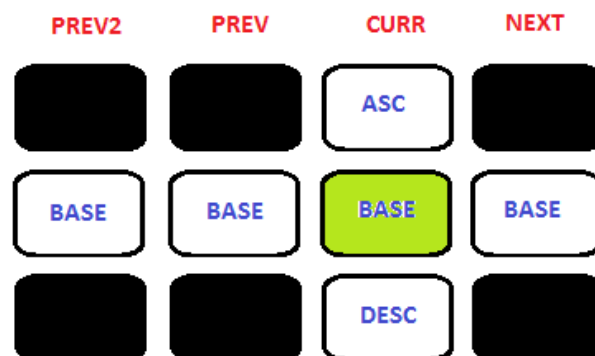


White Box	: Valid
Yellow Box	: 2 Options
Blue Box	: 3 Options

OCR Result and Data Representation

IIA-5-789-P2.jpg_OCR.txt - Notepad

File Edit Format View Help

[illegible]

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{ } arr_line_ocr <3x218 cell>
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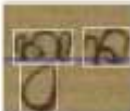
[illegible]

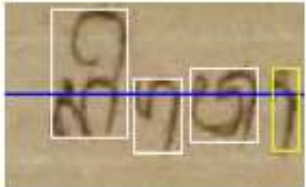

Evaluation: Glyph Segmentation and Recognition (Task 3.2)

Results

No	Manuscript Page	SR	SRR
1	Bangli-P41	49,54	57,94
2	Bangli-P47	58,96	55,70
3	IIA-10-1534-P8	77,37	84,00
4	IIIC-24-1641-P8	77,29	68,56
5	JG-01-P3	70,40	77,28
6	JG-02-P6	30,77	27,50
7	JG-02-P7	40,71	52,63
8	JG-05-P8	60,74	66,67
9	MB-AdiParwa(Purana) -5338.2-IV.a-P30	77,78	77,81
10	MB-AjiGriguh -5783-107.2-P11	30,00	74,07
11	MB-AjiGriguh -5783-107.2-P8	21,83	58,06
12	MB-ArjunaWiwaha -GrantangBasall-P15	64,97	74,87
13	MB-ArjunaWiwaha -GrantangBasall-P28	75,78	71,60
14	MB-TaruPramana-P3	71,11	64,58
15	MB-TaruPramana-P4	43,90	70,05
16	MB-TaruPramana-P6	44,52	54,35
17	WN-P5b	46,06	44,52
18	WN-P7a	11,17	15,00
19	WN-P9a	21,86	42,62

STEP 6: Transliteration with Phonological Rules-based Machine (Task 3.2)

OCR Output	Phonological Rules Output									
 <table><tr><td></td><td></td><td></td></tr><tr><td>'A'</td><td>'KA'</td><td>'BISAH_...'</td></tr><tr><td></td><td></td><td></td></tr></table>				'A'	'KA'	'BISAH_...'				RULE1:... RULE8:...a a RULE1:...K RULE8:...Ka RULE15:...KaH aKaH RULE32:...* aKaH* Final Output: AKAH
'A'	'KA'	'BISAH_...'								
 <table><tr><td></td><td></td><td></td></tr><tr><td>'KA'</td><td></td><td>'NA'</td></tr><tr><td>'SUKU KEMBUNG'</td><td></td><td></td></tr></table>				'KA'		'NA'	'SUKU KEMBUNG'			RULE1:...K RULE2:...KW RULE12:...KWa KWa RULE1:...N RULE8:...Na KwaNa Final Output: KWANA
'KA'		'NA'								
'SUKU KEMBUNG'										

OCR Output	Phonological Rules Output												
 <table><tr><td>[]</td><td>[]</td><td>[]</td><td>[]</td></tr><tr><td>'NI'</td><td>'TALENG'</td><td>'WA'</td><td>'BISAH_...</td></tr><tr><td>[]</td><td>[]</td><td>[]</td><td>[]</td></tr></table>	[]	[]	[]	[]	'NI'	'TALENG'	'WA'	'BISAH_...	[]	[]	[]	[]	RULE17:...NI NI RULE32:...* NI* RULE1:...W RULE3:...WE RULE15:...WEH NI*WEH RULE32:...* NI*WEH* Final Output: NIWEH
[]	[]	[]	[]										
'NI'	'TALENG'	'WA'	'BISAH_...										
[]	[]	[]	[]										
 <table><tr><td>[]</td><td>[]</td><td>[]</td><td>[]</td></tr><tr><td>'TALENG'</td><td>'NA'</td><td>'TEDONG'</td><td>'RA'</td></tr><tr><td>[]</td><td>[]</td><td>[]</td><td>[]</td></tr></table>	[]	[]	[]	[]	'TALENG'	'NA'	'TEDONG'	'RA'	[]	[]	[]	[]	RULE32:...* RULE1:...N RULE4:...No No RULE32:...* No* RULE1:...R RULE8:...Ra No*Ra Final Output: NORA
[]	[]	[]	[]										
'TALENG'	'NA'	'TEDONG'	'RA'										
[]	[]	[]	[]										

Evaluation: Text Line Transliteration (Task 3.2)

Evaluation Metrics

- The text pattern \rightarrow generalized suffix tree [Kesiman, 2006] between two transliterated texts
- Pattern rate (**PR**) = percentage of the same text pattern between the transliteration text and the ground truth text
- Minimal length of pattern text = 4 characters = minimal length of Balinese word with two basic glyphs
- Recall pattern rate (**RPR**) = **PR** / length of ground truth text
- Precision pattern rate (**PPR**) = **PR** / length of the transliteration result

[illegible]

Left : Example of Generalized Suffix Tree between GT string "*madewinduntarakesiman*" and Evaluated string "*malewinduandarakeriman*", Right : the Pattern Tree

Evaluation: Text Line Transliteration (Task 3.2)

MB-AdiParwa(Purana)-5338.2-IV.a-P30.jpg_line_1

GT:73.5484

tadharMANINGATUNG GATUNGga,leGASARATADANASARWAbARANAarTAMASwastramuliADAWALA.yakaTAMADIANINGASARAPATUNGAN.NIS
TASARATUNGA,legADANASARWAbOJARAPAna.uTAMANINGsa

Result:56.3758

wadhewrene0,wa**NGATUNGA**,2**GASARAsADANA**wwabh**ARANA**h**ARWAMAS**umul**ADAWALA**,,yenad**AMADhIANIASARAPA**wa**NGANINI**,jah,**NISTAS**
ARATU2gnabhanayarwuane**JARANA**,,3**TAMANI**sa4a

MB-AdiParwa(Purana)-5338.2-IV.a-P30.jpg_line_2

GT:72.619

RATUNGA,LEGASARASAsAJININGMATI,DINuLURASARwasekARWIJAWIJAN,NINGAsep,yekAUTAMANINGDHARmasuRATUNGA.mUAHNISAPA
MidARATekengTUNON,anruWUSKARSAWASumeNGKAhengPATAWUlan,tiNADI

Result:64.9425

dha5**RATUNGA**,LEGASARASAcAJINIMATI,DINI**LURA**yawua,rng**WIJARWIJAN,NINGA**upapa,,yer**AUTAMAN**idhmamasng**RATUNGA**,,,w**UAHNISAPA**
Mm,DHARATkei**TUNON**,jnsatua**WUSKASAWASaNGKA**ie**PATAWU**panr,tar**NADI**

MB-AdiParwa(Purana)-5338.2-IV.a-P30.jpg_line_3

GT:71.5232

nan.madiANINGAMIDARA,RihUWUSNINGsawATINUNU,TikaADIN.uTAMANINGAMIDARA,brasTATAANAWUTikangsaWA,TinADINAN.yeKATATAk
ausAPARIPAMIDANGAN,muangKAUCAPATATABASA

Result:60.2564

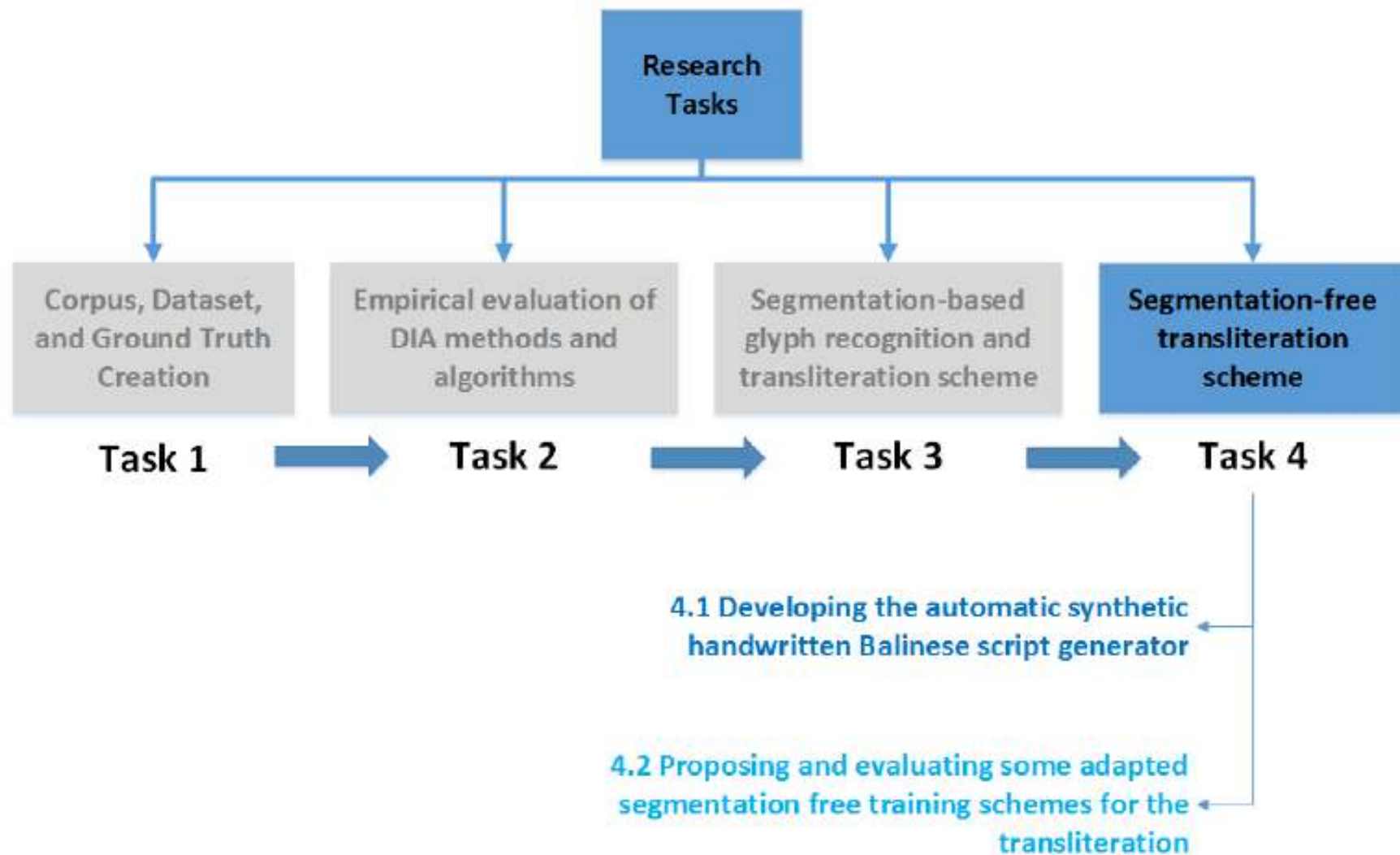
5nrabamadhwa**ANINGA**jmwidh**ARA,RIUWUS**uwu**ATINUNU**,Tark**ADINA**ca,,wa**TAMANIAMID**haraung4staj**ATAANAWUT**lisng**WA**,Tunnn**INAN**,,pena
KATATAKAUCAPARIPAMadh**ANGA**,wi**KAUCAPATATABASA**

The text pattern (in red) extracted between ground truth text and result text

Evaluation: Text Line Transliteration (Task 3.2)

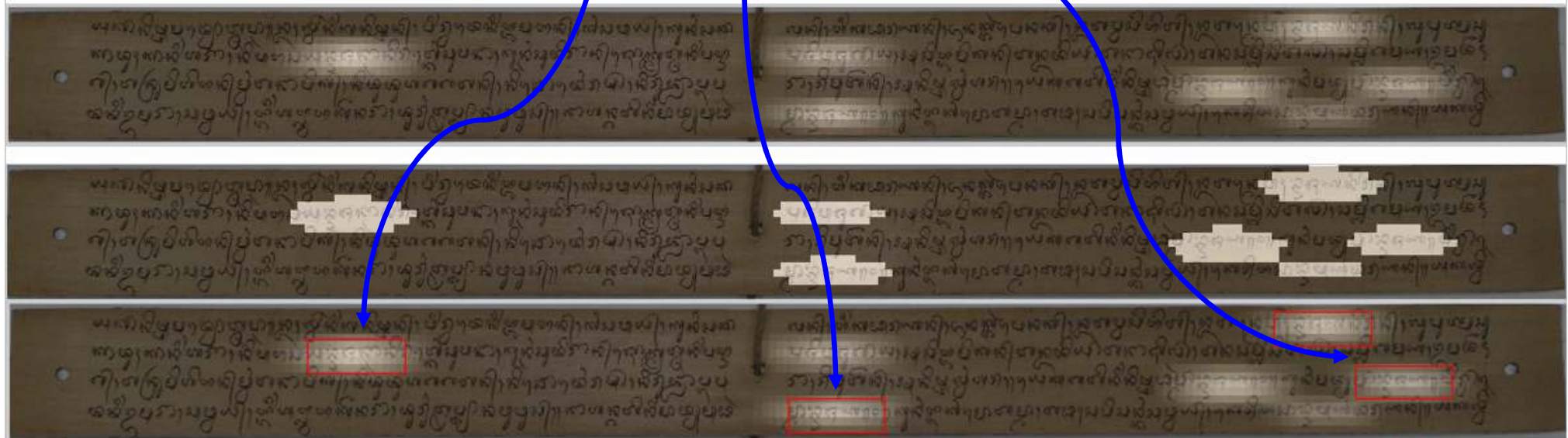
No	Manuscript Collection	RPR		PPR	
		Max	Avg	Max	Avg
1	Bangli	69,23	6,26	53,13	7,36
2	IIA-10-1534	73,13	47,78	68,16	43,73
3	IIA-5-789	76,25	50,27	77,85	48,67
4	IIB-2-180	72,63	54,04	69,61	52,69
5	IIIB-12-306	67,78	34,84	64,91	33,43
6	IIIB-42-1526	78,61	51,50	71,43	50,06
7	IIIB-45-2296	65,09	40,29	65,61	39,15
8	IIIC-19-1293	60,38	35,52	63,40	36,56
9	IIIC-20-1397	75,58	34,44	60,48	34,46
10	IIIC-23-1506	54,46	27,98	56,82	25,02
11	IIIC-24-1641	60,44	40,61	54,49	41,91
12	JG-01	56,47	23,88	57,72	20,70
13	JG-02	11,11	0,80	22,86	1,53
14	JG-03	25,83	4,91	23,88	7,34
15	JG-04	14,49	1,37	15,38	1,64
16	JG-05	27,37	4,55	41,94	8,85
17	JG-06	0,00	0,00	0,00	0,00
18	JG-07	4,94	0,15	12,90	0,39
19	MB-AdiParwa(Purana) -5338.2-IV.a	83,13	37,43	72,44	36,54
20	MB-AjiGriguh -5783-107.2	4,71	0,28	10,81	0,64
21	MB-ArjunaWiwaha -GrantangBasall	65,81	26,83	68,18	32,89
22	MB-TaruPramana	70,15	17,69	66,67	18,99
23	WN	20,00	0,72	10,53	0,68
	All: 390 pages				

Segmentation-free Transliteration Scheme

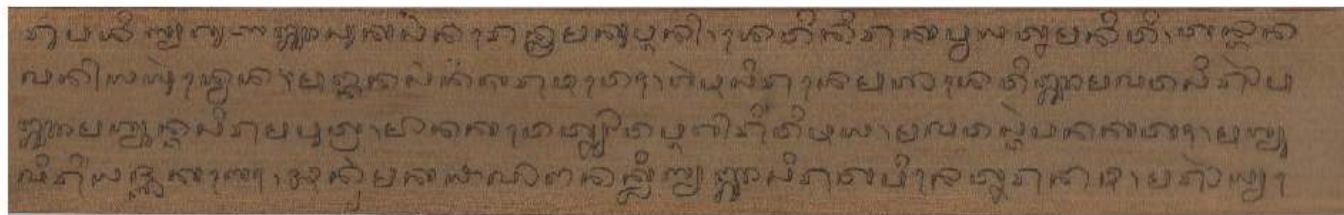


Towards Word Spotting System

citra query ("bhujana")



Word Spotting Challenges



3 cm

- One leaf = 2 pages
- One page = 4 text lines
- Number of leaves in one collection, example :
 - ✓ Collection of **Kakawin Ramayana** - **07.41/5634.3** = 140 leaves
 - ✓ Collection of **Kakawin Ramayana** - **07.40/5634.3** = 155 leaves

❑ The segmentation-based word spotting:

- ✓ not trivial to be applied
- ✓ no spaces between each word.
- ✓ the use of ascender, descender and conjunct form: it can be written jointly with the previous word.

❑ The segmentation-free word spotting method:

- ✓ to extract all possible word patches in the manuscript pages: Offline Patch Images Extraction
- ✓ to detect and to separate possible text area from the textured image of palm leaf
 - ❖ cover all text areas in manuscript page
 - ❖ spot as close as possible all words in manuscript
- ✓ to reduce the amount of processing time,
- ✓ to ignore many false patch image positions: non text area and the area between the unstraight textlines.

In this paper, we present our proposed segmentation-free word spotting method for the Balinese palm leaf manuscripts.

ကျွန်းကြွယ်ပေါ်မှာရှိတဲ့အားကိုးပါးကိုးပါး

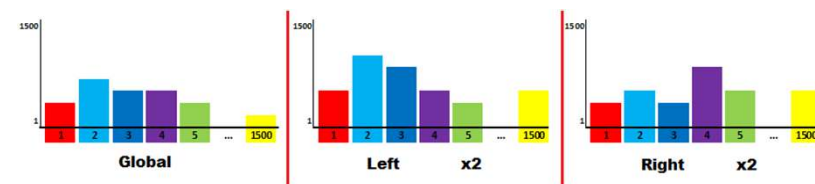
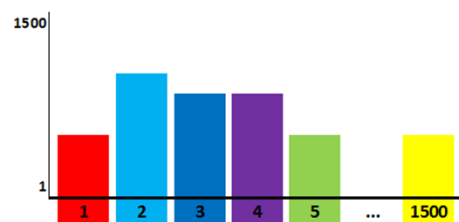


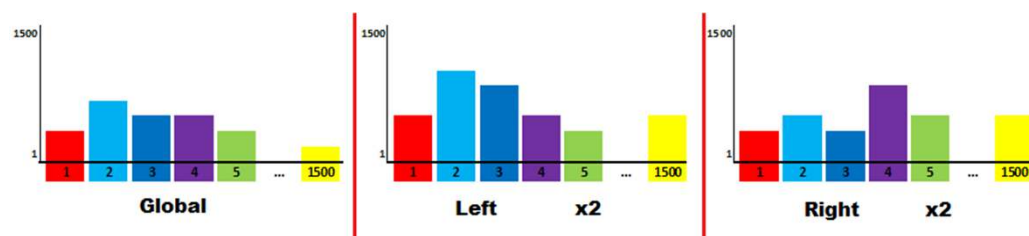
ကျွန်းကြွယ်ပေါ်မှာရှိတဲ့အားကိုးပါးကိုးပါး

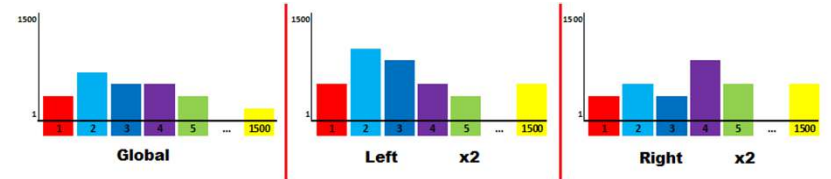
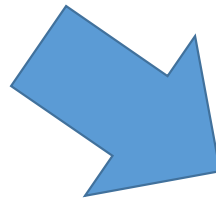
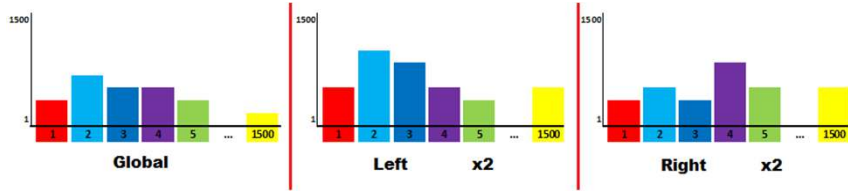
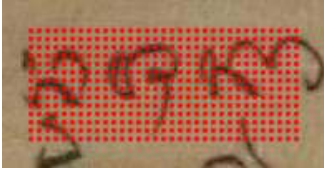


ကျွန်းကြွယ်ပေါ်မှာရှိတဲ့အားကိုးပါးကိုးပါး

Visual Word



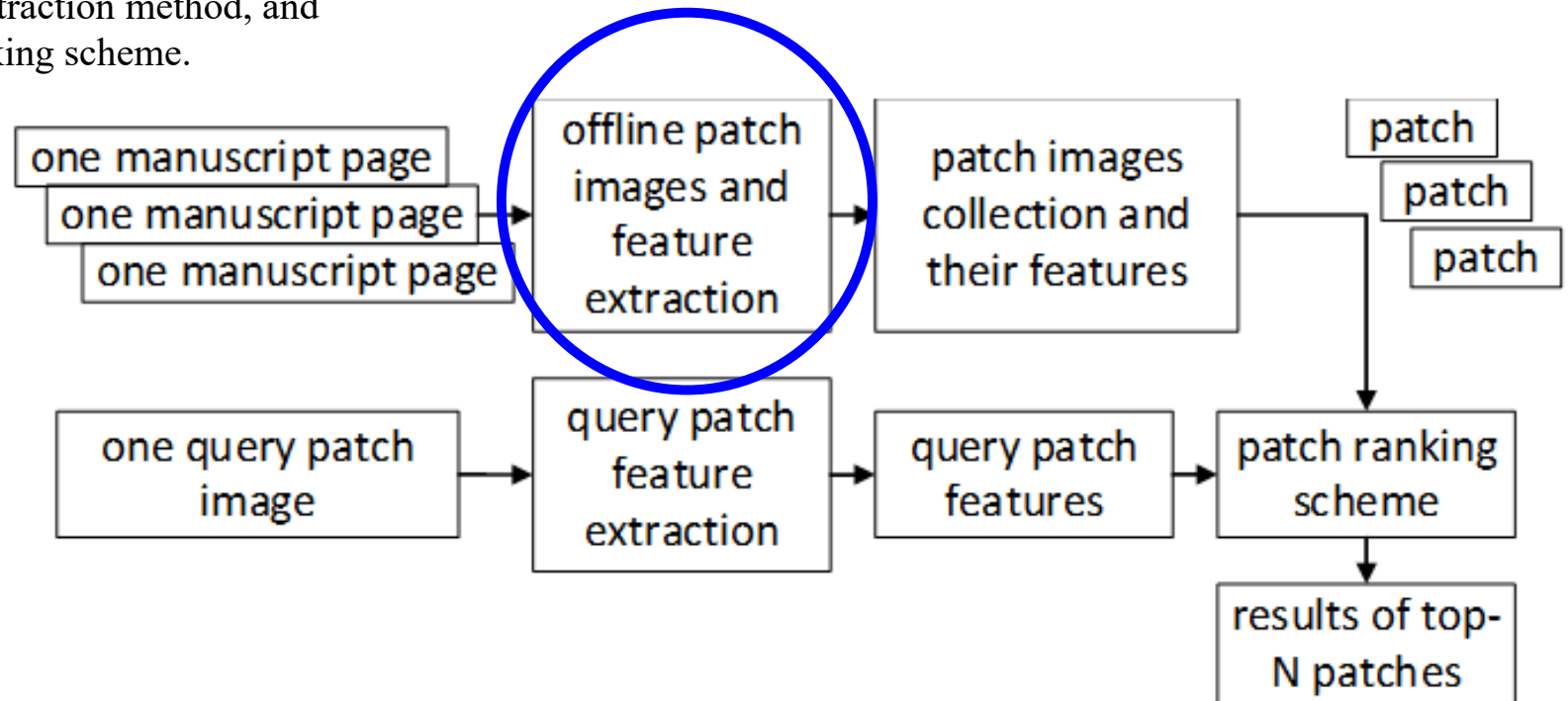




Proposed Scheme : Global Workflow of Word Spotting Scheme

Three main sub schemes:

- ✓ the offline patch images extraction process,
- ✓ the feature extraction method, and
- ✓ the patch ranking scheme.



Challenges in Word Patch Extraction

- to detect the text area of the degraded image of Lontar with many different text layouts
 - ❖ only full text, contain graphics with large blank area, the text is written in table-like format
- to cover all text area: optimally avoiding the blank area
- to reduce the possible number of word patches: less number of extracted word patches, the word spotting system will perform faster
- the text lines are not always written in straight position from left to right



Fig. 4. Lontar page with full text

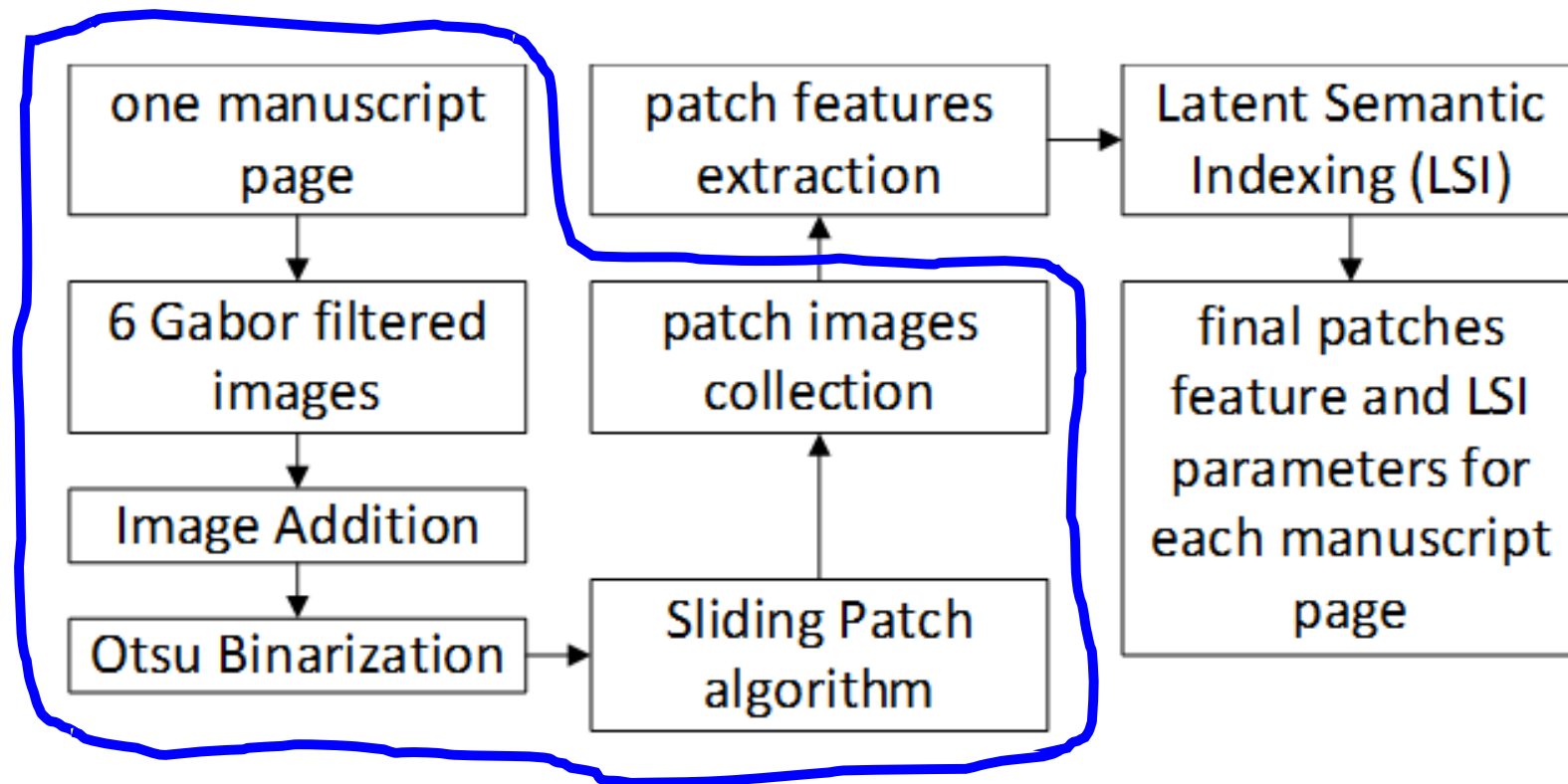


Fig. 5. Lontar page with blank area



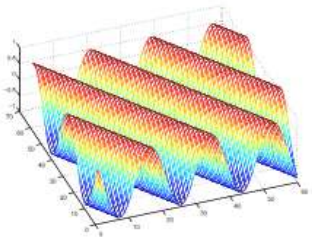
Fig. 6. Lontar page with table-like format

Offline Patch Images Extraction Process

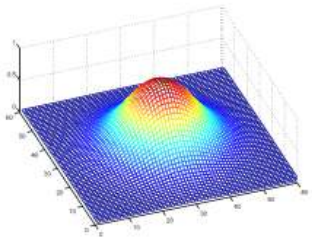


Gabor Filters

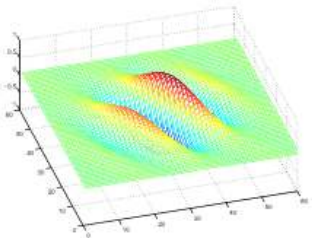
Prasad, V.S., & Domke, J. (2005). Gabor Filter Visualization.



(a)



(b)



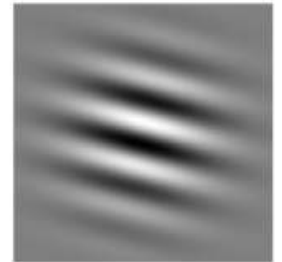
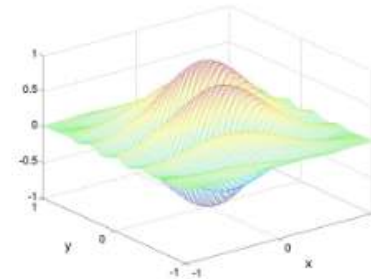
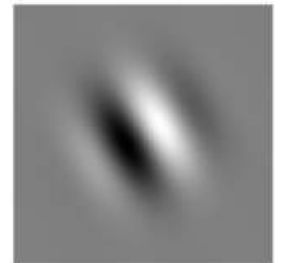
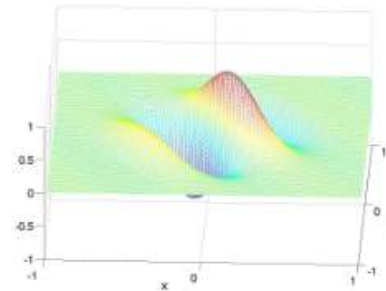
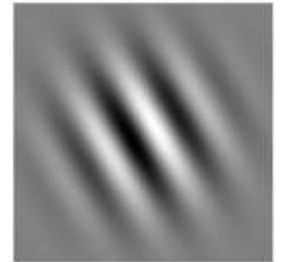
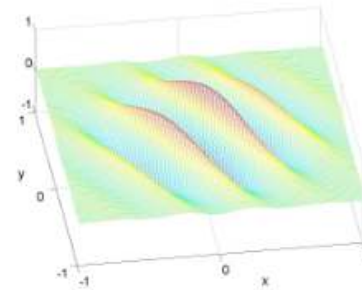
(c)

✓ Lontar:

- The writing: shows the spatial texture informations.
- The different writers: different frequencies and orientations of the textures.

✓ Gabor filter:

- Modulation between sinusoid and Gaussian filter.
- A texture filter with many orientations and frequencies
- Bank of Gabor filters : orientation, wavelength, aspect ratio and bandwidth
- Provide initial information about the existence of textures in the document.
- Detect the preliminary informations about text and non text area on Lontar.



Gabor filter composition: (a) 2D sinusoid oriented at 30° with the x-axis, (b) a Gaussian kernel, (c) the corresponding Gabor filter. Notice how the sinusoid becomes spatially localized.

Example of Gabor filters with different frequencies and orientations. First column shows their 3D plots and the second one, the intensity plots of their amplitude along the image plane.

Gabor-filtered Images



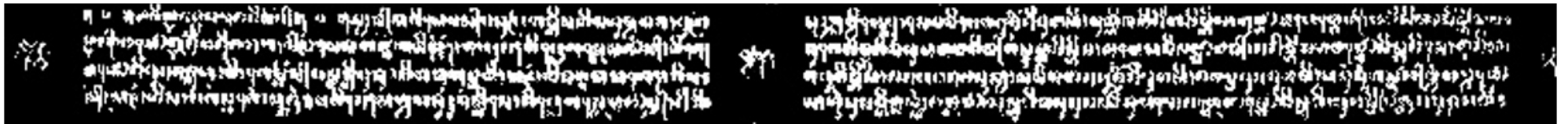
- Six different orientations: 0° , 45° , 135° , 180° , 225° , and 315° .
- Wavelength: 8,
- Aspect ratio: 0.5,
- Bandwidth: 1.

The result of six Gabor-filtered images will be joined (with a simple image addition operation) to produce one grayscale Gabor image.

Six Gabor-filtered images of a manuscript page

The Binary Mask Image

Next step: Binarization with the global Otsu's : a binary mask image for the adaptive sliding patch algorithm.



The binary mask image for the adaptive sliding patch algorithm

Optimized Sliding Window Algorithm

To optimally extract the image patches only in the text line area of the Gabor filtered images

The algorithm :

1. A sliding patch of size 125 (height) x 300 (width) pixels will be moved (from left to right with 100 pixels step and up to down with 50 pixels step)
 - ✓ Will be spotted many patch images in manuscript page.
2. On each spotted patch position :
 - ✓ calculate the ratio (R) of the number of white pixels (text) and the total number of pixels in patch image.
 - ✓ calculate the number of white pixels (text) only in upper (U), middle (M) and lower (U) part of the patch image.
3. A conditional rule is then applied :
 - ✓ whether this patch image contains a significant number of text area (when $R > 0.1$)
 - ✓ and a good centered-spotting position on the manuscript (when $M > U$ and $M > L$).
4. If the condition is met : save the patch image position.
5. Otherwise : move the sliding patch 2 pixels in a lower vertical position repetitively until the condition is met.
 - ✓ It should be noted that during this repetitive patch searching in vertical position, the horizontal position of the patch is kept.

Adaptive Sliding Patch Algorithm

Input: IMG_MASK : binary image as mask
IMG : manuscript gray scale image
Output: all possible word patch images

Algorithm:

```
h=125 (sliding patch height)
w=300 (sliding patch width)
h_from=1; h_to=h_from+h-1; (sliding patch height position)

while (h_from and h_to are still inside height image)
    w_from=1; w_to=w_from+w-1; (sliding patch width position)
    allow_slide=0;

    while (w_from and w_to are still inside width image AND
           h_from and h_to are still inside height image)

        Get PATCH from IMG
        Get PATCH_MASK from IMG_MASK
        Calculate R of PATCH_MASK
        Calculate U of PATCH_MASK
        Calculate M of PATCH_MASK
        Calculate L of PATCH_MASK

        if (R>0.1 and M>U and M>L) (condition is met)
            Extract this PATCH
            if (allow_slide==0)
                h_from=h_from_save;
                h_to=h_to_save;
                allow_slide=0;
            end (of if)
        else (condition is not met)
            if (allow_slide==0) (remember the patch height position)
                h_from_save=h_from;
                h_to_save=h_to;
            end (of if)
            if (allow_slide<10) (maximum slide 10 times)
                h_from=h_from+2; h_to=h_to+2;
                w_from=w_from-w_step; w_to=w_to-w_step;
                allow_slide=allow_slide+1;
            else (no more slide, back to the saved patch height position)
                h_from=h_from_save;
                h_to=h_to_save;
                allow_slide=0;
            end (of if-else)
        end (of while)

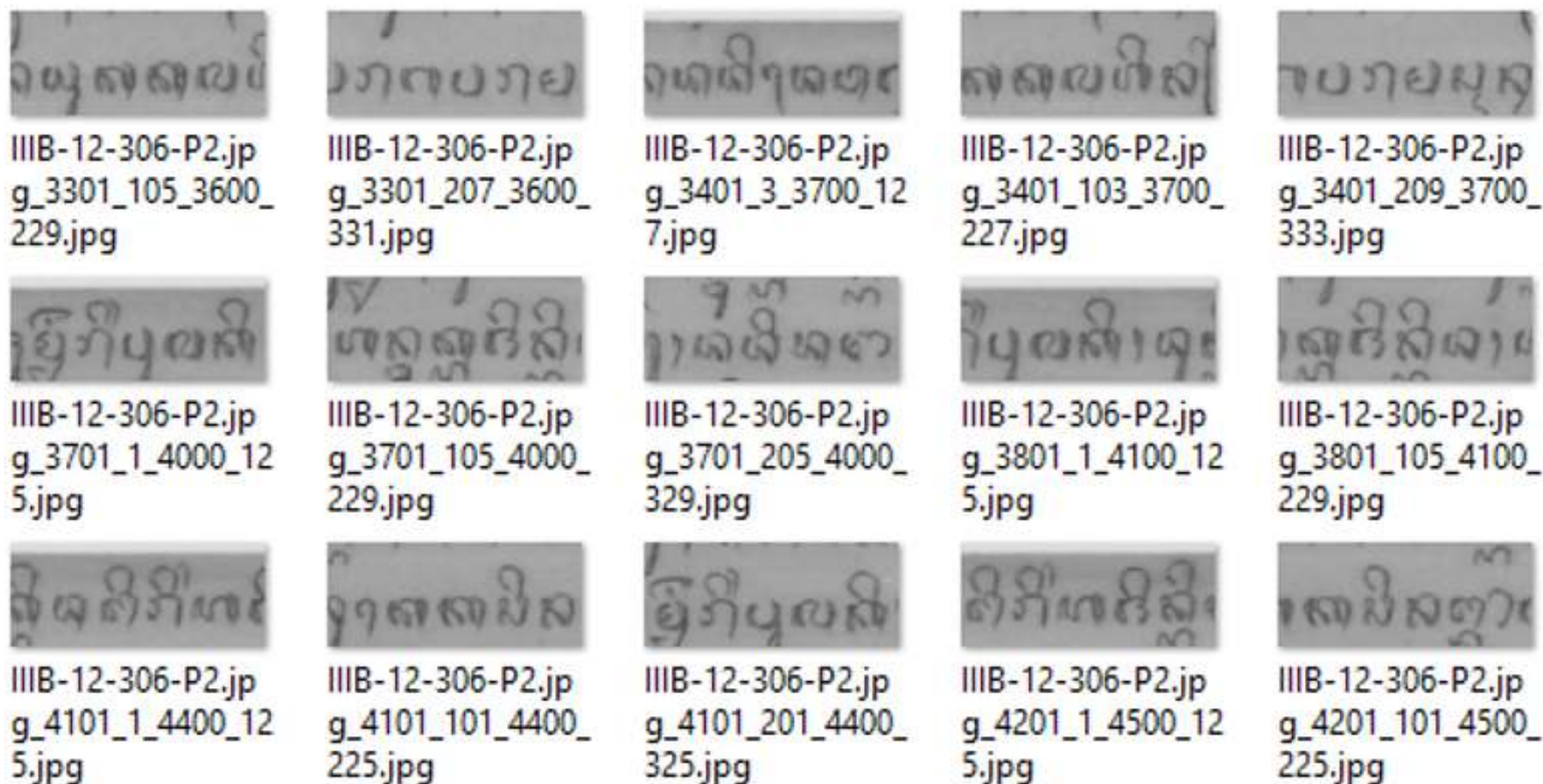
        w_from=w_from+w_step; w_to=w_to+w_step;
    end (of while)
    h_from=h_from+h_step; h_to=h_to+h_step;
end (of while)
```

Experimental Results



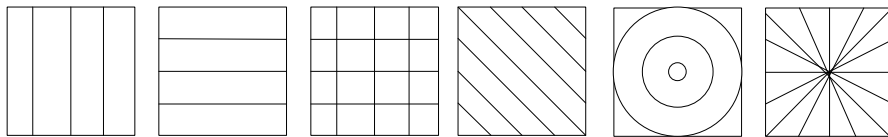
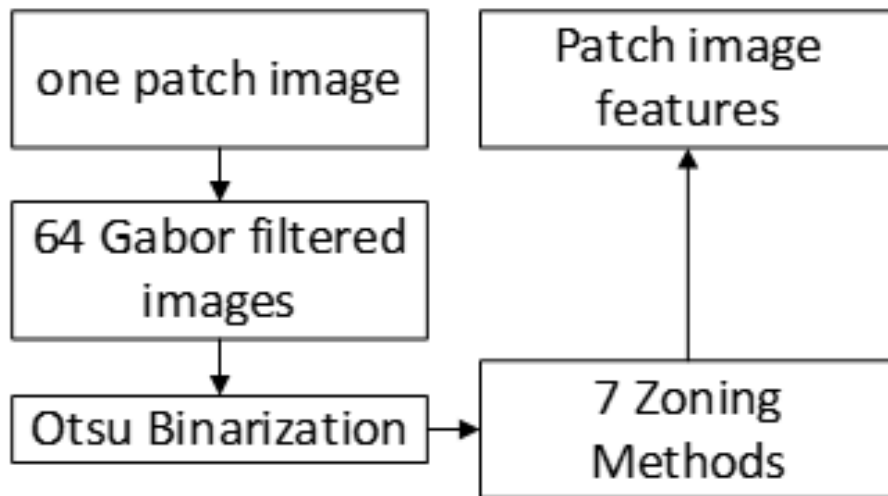
Examples of text patch area extraction from the manuscript pages

- ❑ The combination of Gabor filter with the optimized sliding window algorithm is effectively able to detect and to extract image patches from the text area of the Balinese palm leaf manuscripts
- ❑ This procedure is well adapted to cope with the unstraight text line in manuscript.



Some examples of the extracted patch images from a manuscript page

Feature Extraction Method



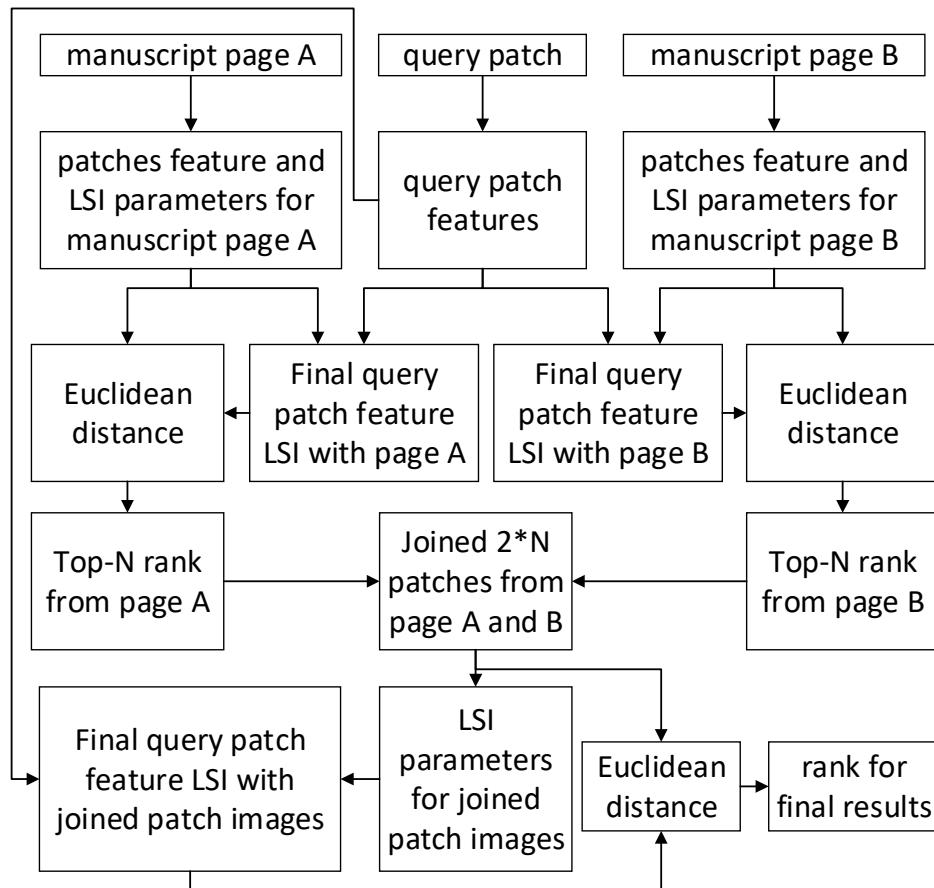
The different Zoning methods

64 Gabor filters + seven different Zoning methods.

A patch image:

- Preprocessed by resizing into 100 x 200 pixels.
- 64 different combinations of Gabor filter parameters
 - 0°, 45°, 90°, 135°, 180°, 225°, 270°, and 315° for the orientation,
 - 4 and 8 for the wavelength,
 - 0.5 and 0.9 for the aspect ratio, and
 - 0.5 and 1 for the bandwidth.
- Global Otsu's binarization method.
- Seven Zoning methods: (distance between zone area is 50 pixels)
 - vertical, horizontal, block, left diagonal,
 - right diagonal, radial and circular
- Feature value: ratio of the number of white pixels (text) and the total number of pixels in zoning area
- Produce the total of 2,240 feature dimensions for each patch image.

Patch Ranking Scheme



We are not directly rank the similarity (or it can be the dissimilarity) between the query patch features and the patch image features all together. However, we first rank the patch image features page per page.

The original patch image feature values and the original query patch feature values will be mapped into new feature values with new feature dimensions by using the Latent Semantic Indexing concept with Single Value Decomposition [9].

In the first LSI step, the feature dimensions will be reduced depending on the number of patches in each manuscript page ($0.5 * nb_patch$). And in the second LSI step, the feature dimensions will be reduced depending on the number of pages in all manuscript collections (nb_page).

Experiments and Results: Dataset and Evaluation

A published dataset of AMADI_LontarSet [10]:

- ✓ Already proposed for word spotting challenge in ICFHR 2016 competition [6].
- ✓ For query-by-example word spotting dataset, it consists of 100 manuscript pages and 36 query patch images for test and evaluation.

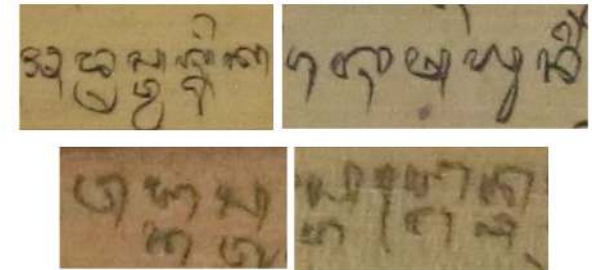


Fig. 13 Some examples of query patch images (“adrestwasnika”, “sesayuning”, “wangkas” and “anggen”)

- Two evaluation measures: mean Recall (mR) and mean average Precision (maP) [1], [5], [9], [11].
- Based on the number of relevant patch spotting area from the spotting results.
- A relevant spotting area: overlapped in more than 30% patch area and the height and width of the spotting patch area are not twice bigger than the height and width of the ground truth patch area.

Experiments and Results

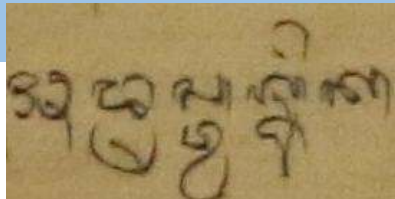


Fig. 14. Relevant spotting for query “*adrestwasnika*”

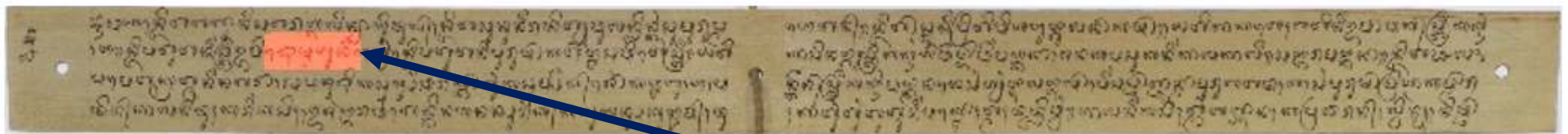
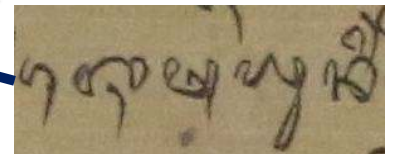


Fig. 15. Relevant spotting for query “*sesayuning*”



Experiments and Results



“wangkas”

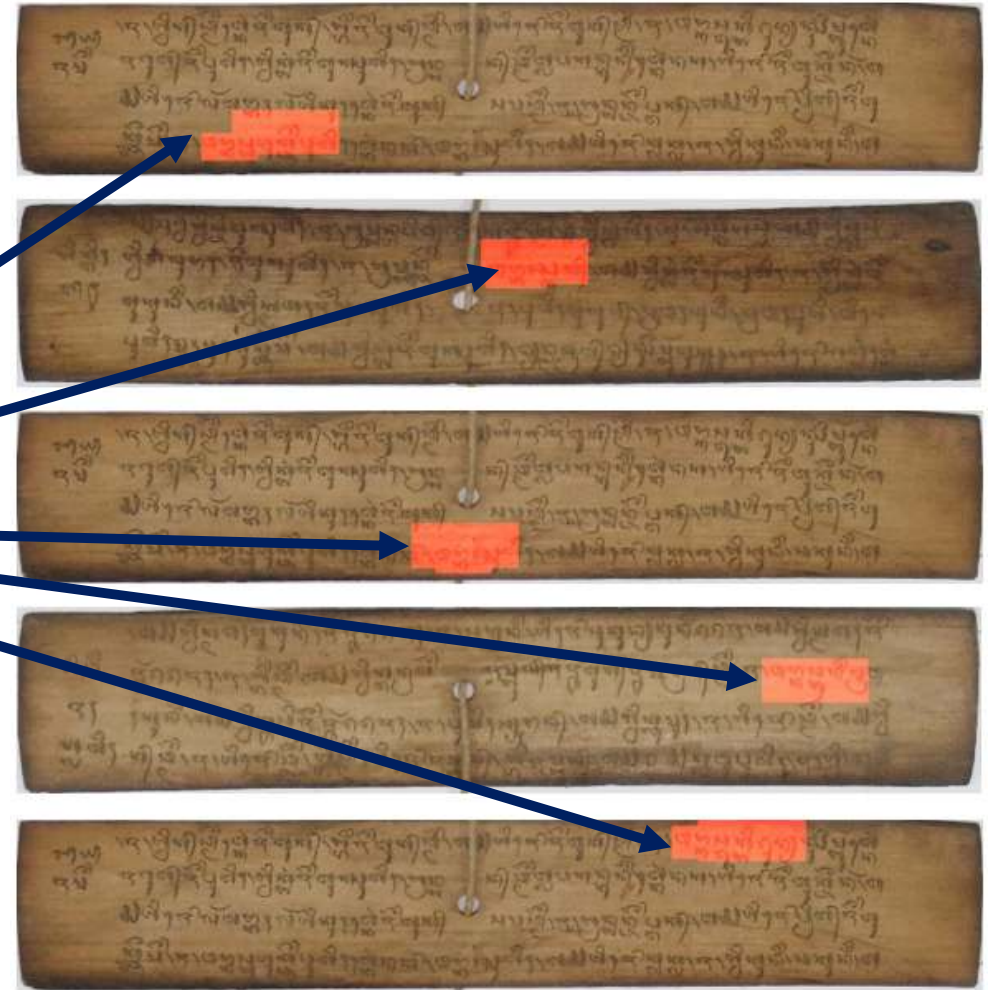


Fig. 16. Five relevant spotting for query “wangkas”

Experiments and Results

1. Achieve a high Recall value for longer query words.
2. This finding can be very useful to define an optimal patch size in patch extraction algorithm for the futur works.
3. The intra-similarity between patches from the same manuscript page is stronger than the intra-similarity between patches from the same query word image.
4. Our proposed patch ranking scheme is able to rank hierarchically all patches from the same page, but still shows low performance in ranking the patches from different pages. It makes the mean average Precision (maP) values are low.

Query Patch Image	Nb Patch		Recall	maP
	Ground Truth	Relevant Spot		
adrestwasnika_query1.jpg	1	1	100,00	0,19
agung_query2.jpg	10	1	10,00	6,50
akah_query3.jpg	19	5	26,32	0,50
ahungguh_query4.jpg	2	1	50,00	0,06
anggen_query5.jpg	22	14	63,64	1,01
banten_query6.jpg	7	4	57,14	0,47
banyu_query7.jpg	4	2	50,00	0,34
candra_query8.jpg	4	0	0,00	0,00
dharna_query9.jpg	14	0	0,00	0,00
dina_query10.jpg	6	0	0,00	0,00
dumlada_query11.jpg	14	1	7,14	0,30
dwijendra_query12.jpg	1	0	0,00	0,00
paturunan_query13.jpg	3	1	33,33	0,13
pawikretan_query14.jpg	2	1	50,00	0,08
sarwa_query15.jpg	13	0	0,00	0,00
sekar_query16.jpg	1	0	0,00	0,00
sesayuning_query17.jpg	1	1	100,00	0,54
siwa_query18.jpg	11	0	0,00	0,00
sri_query19.jpg	9	0	0,00	0,00
stri_query20.jpg	12	0	0,00	0,00
suara_query21.jpg	9	0	0,00	0,00
sudamala_query22.jpg	3	0	0,00	0,00
sunia_query23.jpg	7	0	0,00	0,00
susuhunan_query24.jpg	3	1	33,33	0,06
tadanganing_query25.jpg	1	1	100,00	0,39
taman_query26.jpg	4	1	25,00	0,07
titiange_query27.jpg	14	10	71,43	0,94
tri_query28.jpg	9	0	0,00	0,00
tuan_query29.jpg	9	2	22,22	0,19
tunggal_query30.jpg	17	11	64,71	0,55
wangkas_query31.jpg	5	5	100,00	0,48
widiadara_query32.jpg	3	1	33,33	0,12
widiadari_query33.jpg	3	1	33,33	2,56
windu_query34.jpg	6	2	33,33	0,25
yogiswara_query35.jpg	4	1	25,00	0,13
yuarantaraning_query36.jpg	4	1	25,00	0,07
		mean	30,95	0,44



Thank you