



# Digital Historical Forensics: A Computational Approach to Wartime Media Cultures

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QTM SPEAKER SERIES, EMORY UNIVERSITY

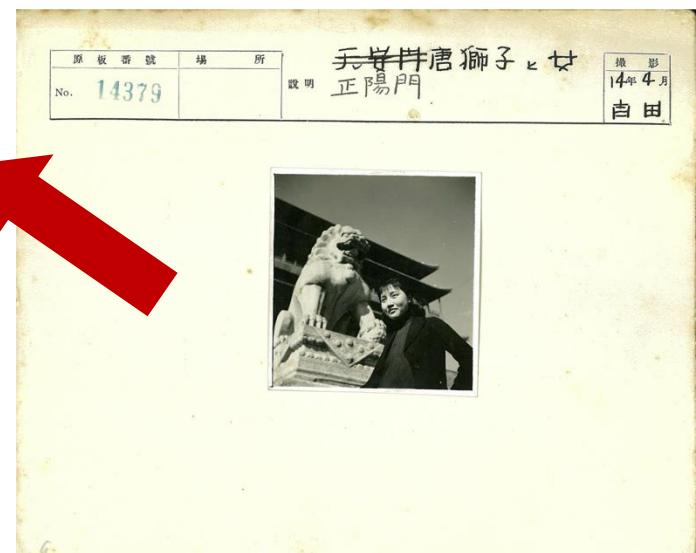
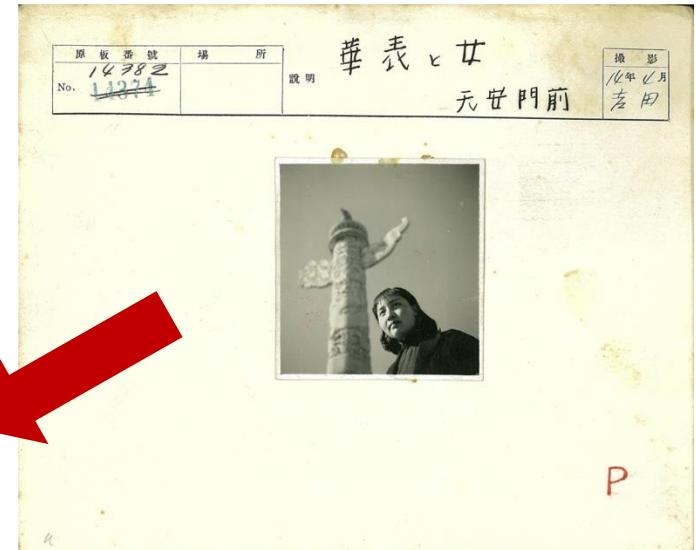
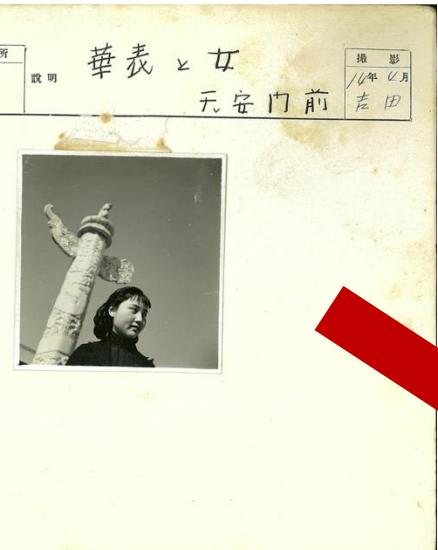
FEBRUARY 24, 2025

# Digital Historical Forensics

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- ❖ This method blends AI technologies—including machine learning and computer vision—with traditional humanities methods, such as close reading and contextual analysis.
- ❖ It bridges empirical precision with interpretive inquiry.





# Digital Historical Forensics

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- ❖ This scalable, data-assisted approach integrates AI with traditional humanities methods, such as close reading and contextual analysis, to enhance media studies.
- ❖ Uncovers meaningful evidence in visual materials, transforming photographs into tools for accessing their historical indexical reality.
- ❖ Sees beyond static photographs and understand the dynamic historical forces at play.

# Challenges and Opportunities in Analyzing Print Media Images

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Dissemination of tons of photographs in 20th-century print media



Challenges to humanities scholars tracing image circulation



The combination of the traditional method—contextual analysis—in media studies and computer vision technologies

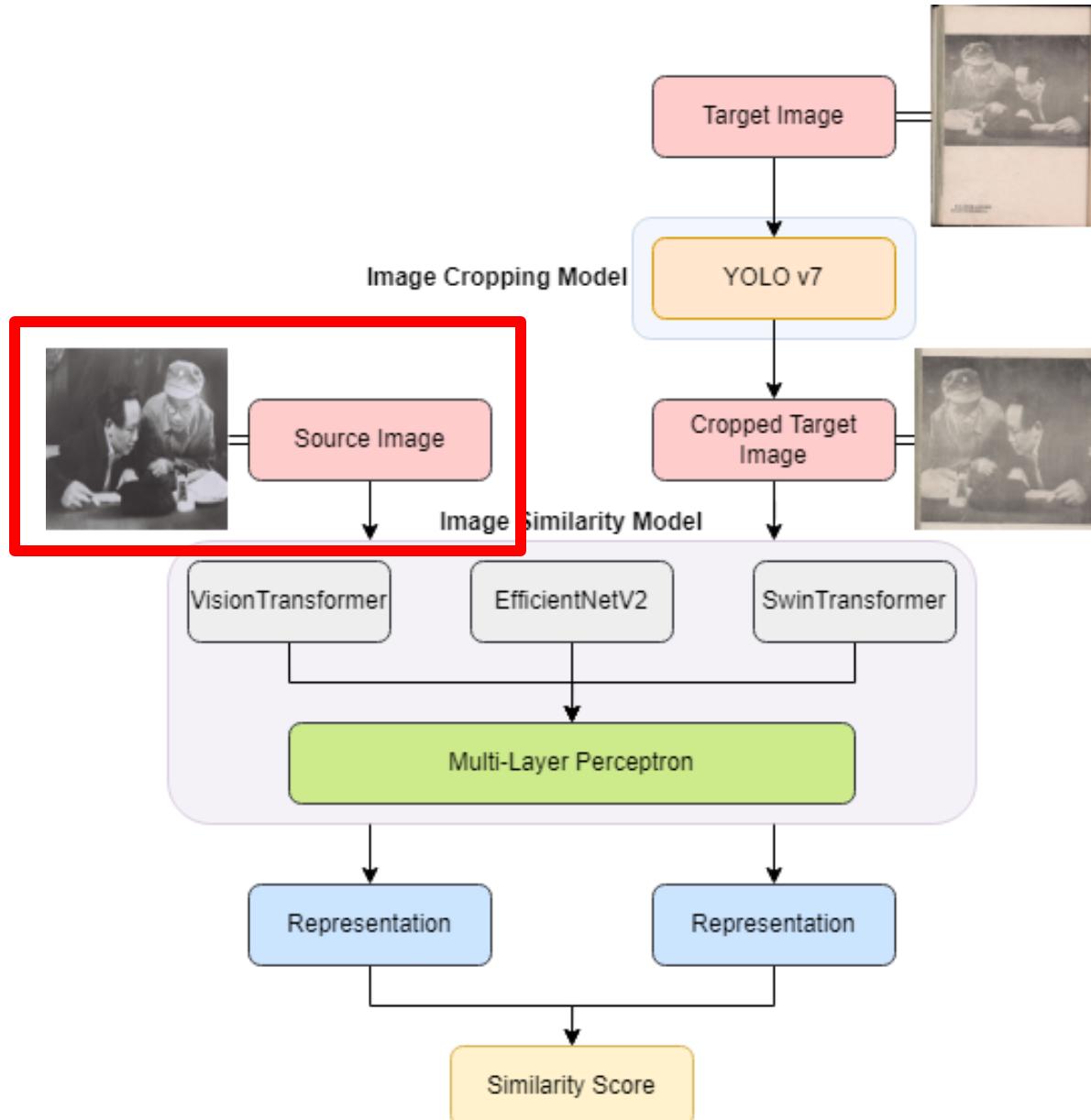
# Case Study: Jinchaji Pictorial series, a significant WWII-era photographic publication of the Chinese Communist Party (1942-1948)

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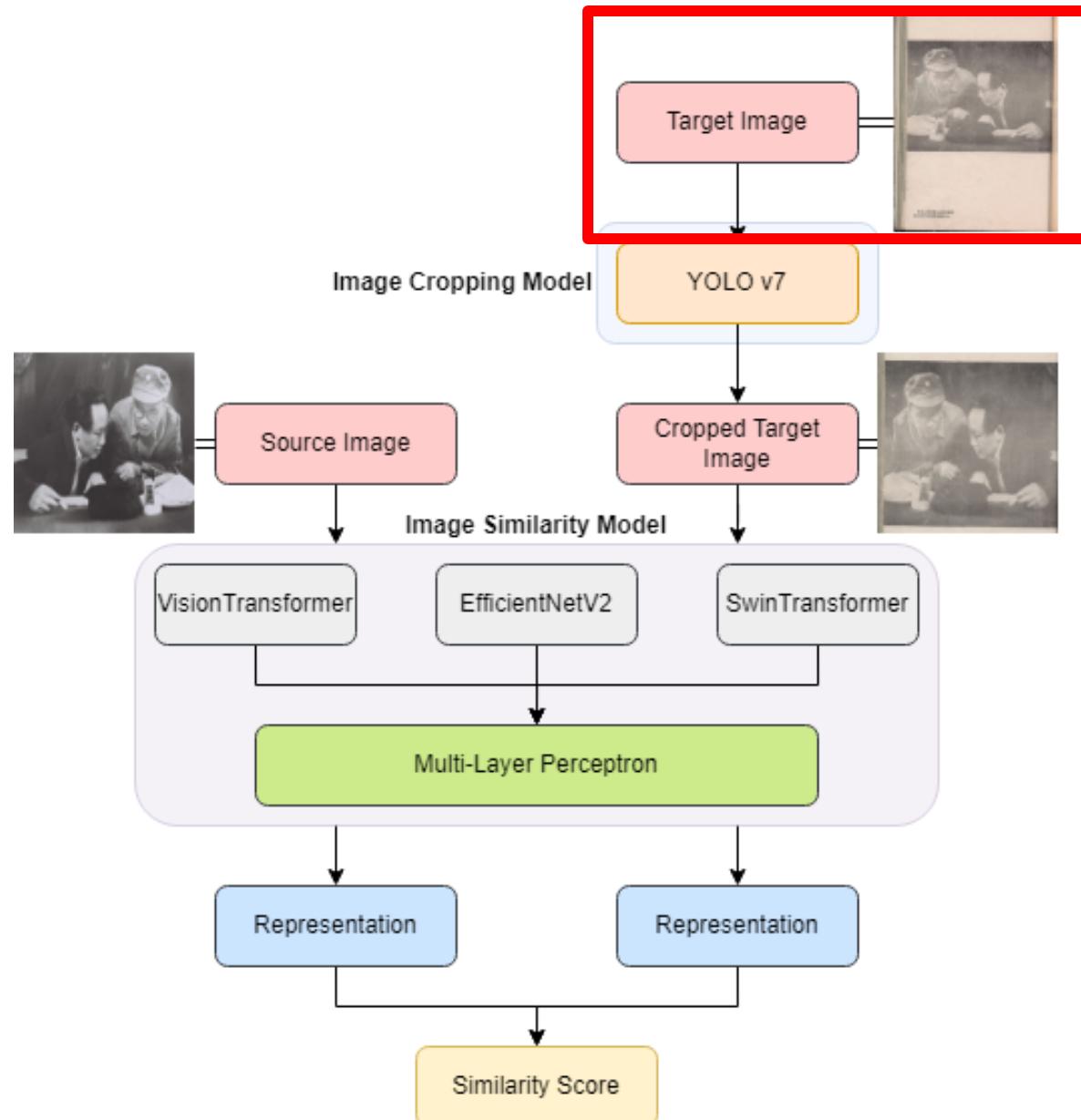
# Computer Vision Pipeline for Image Retrieval

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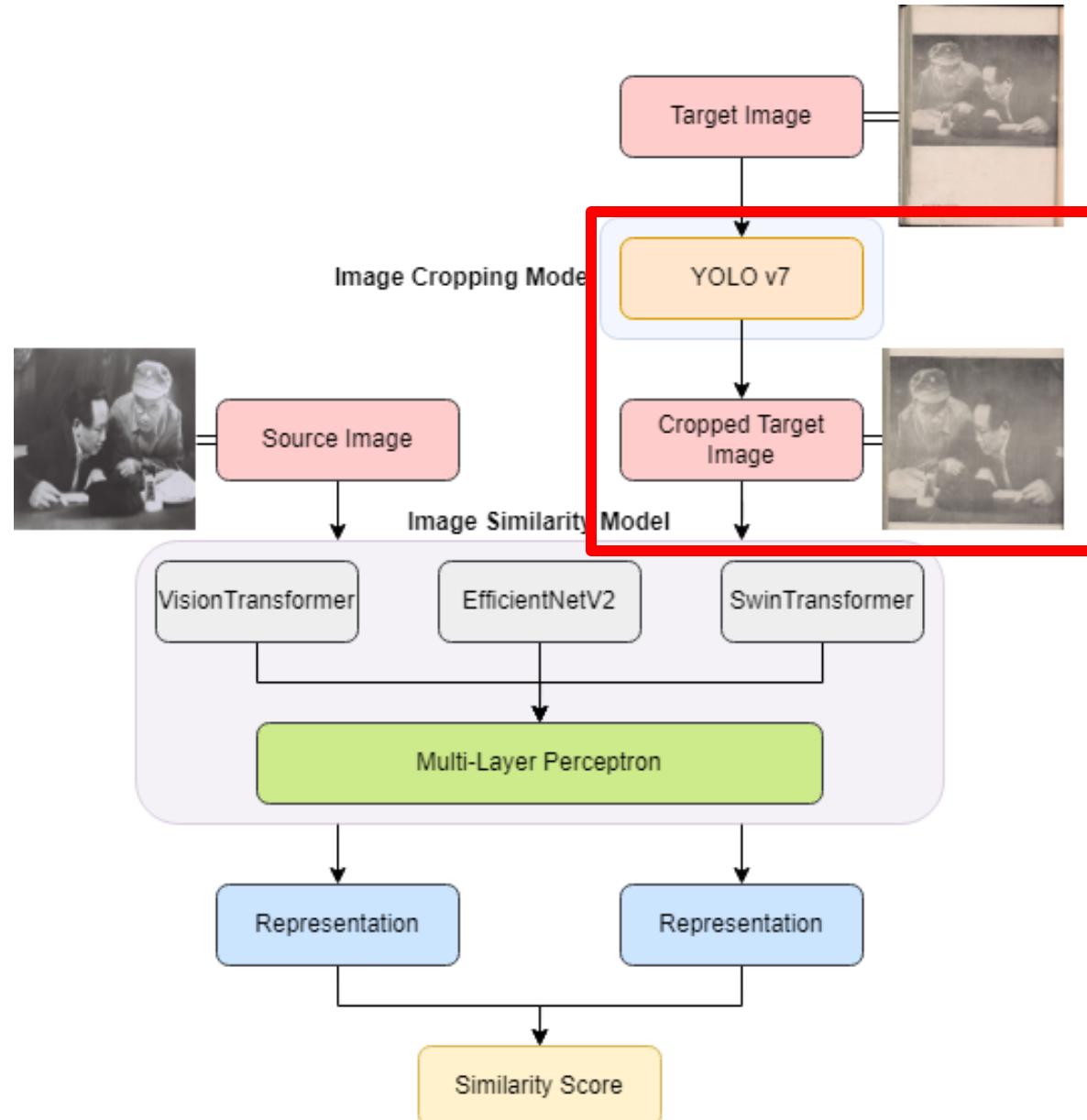
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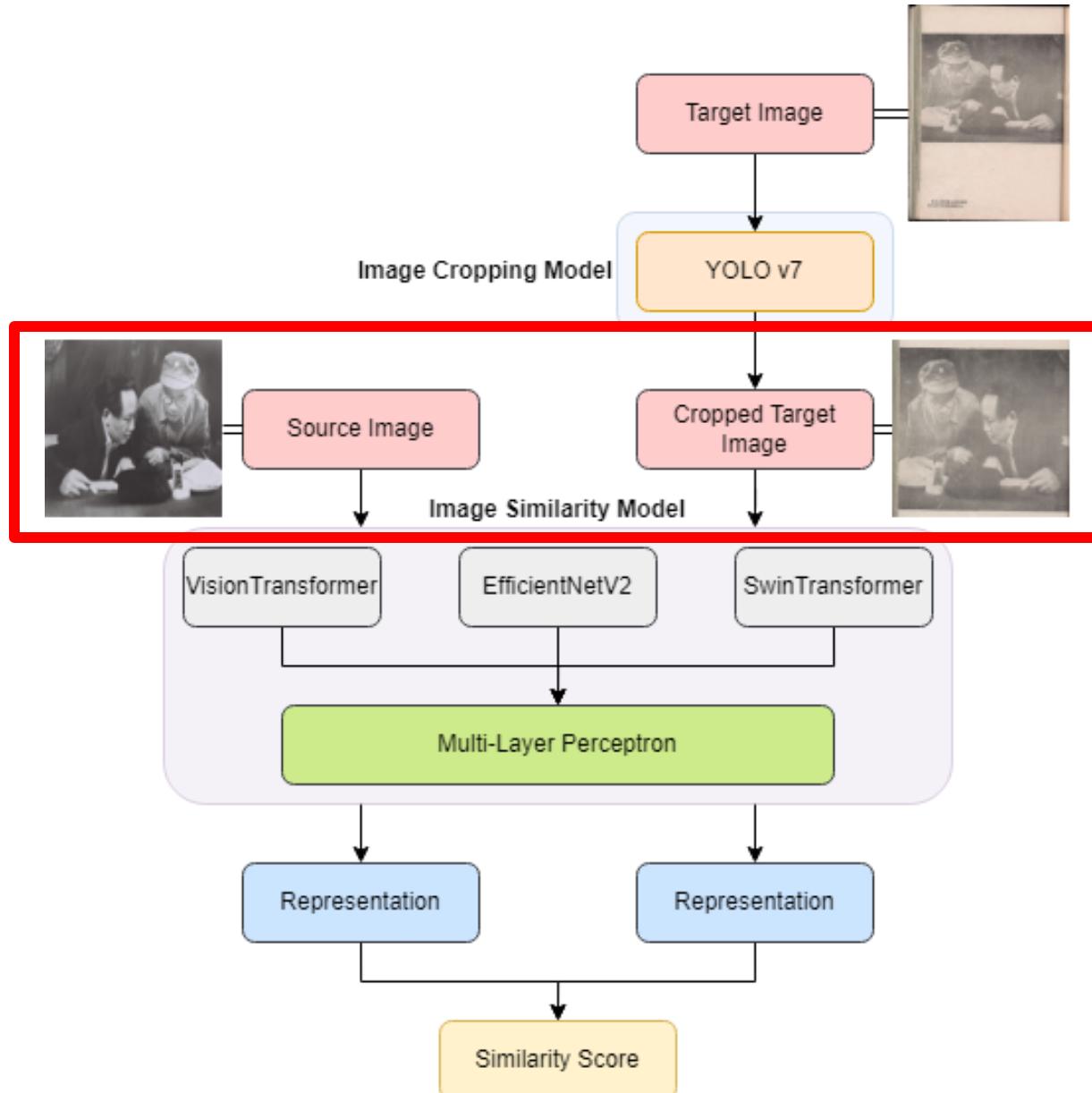
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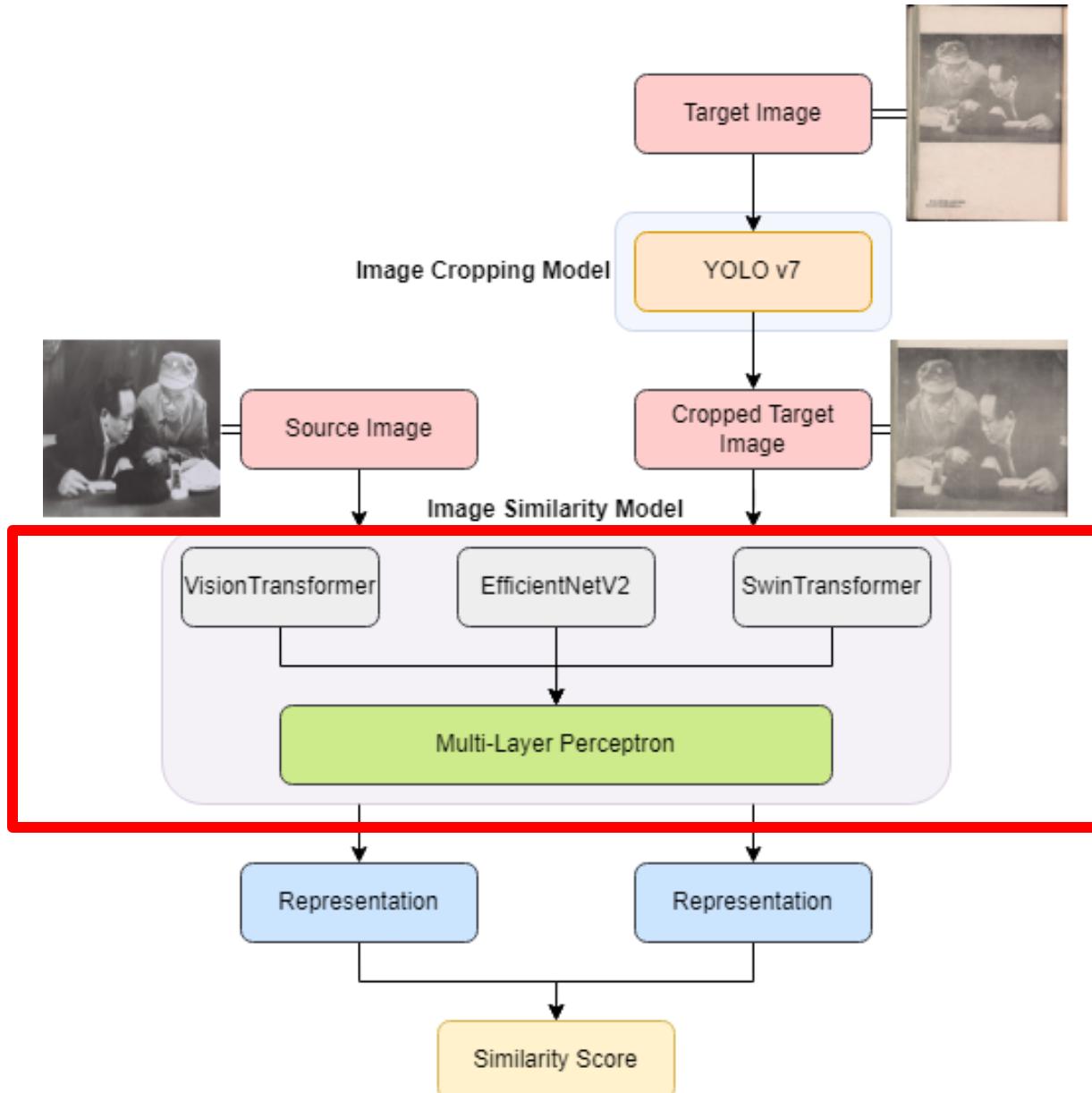
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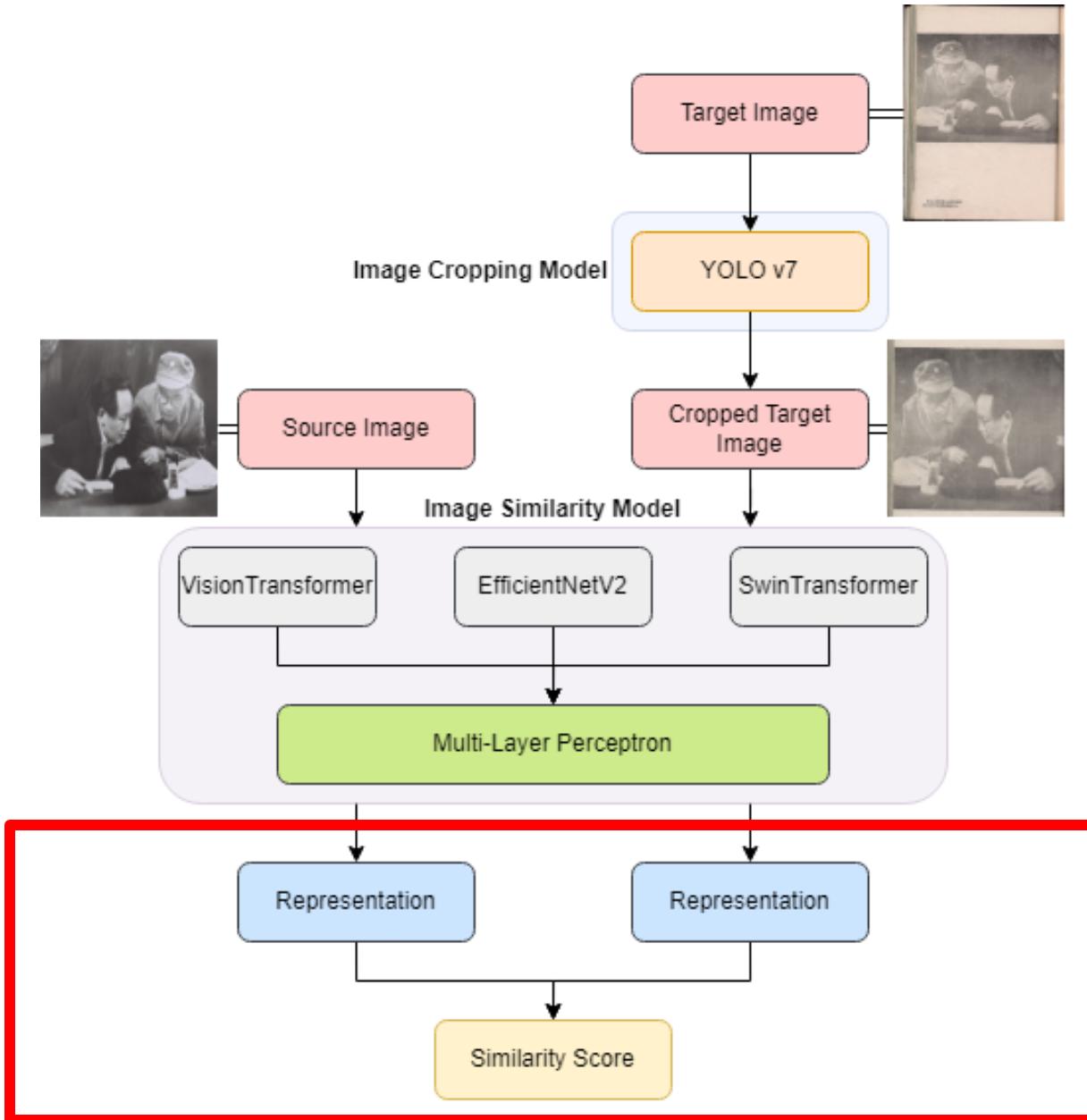
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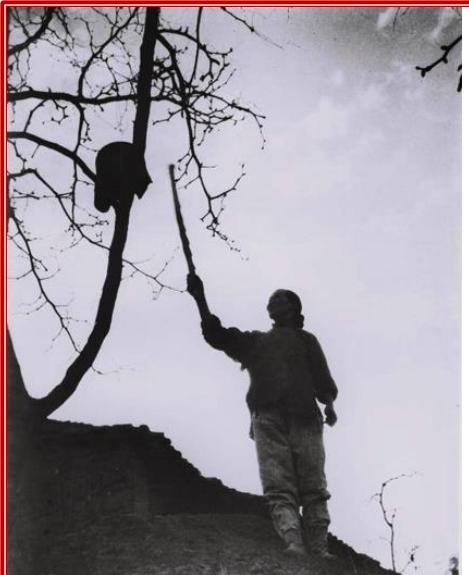
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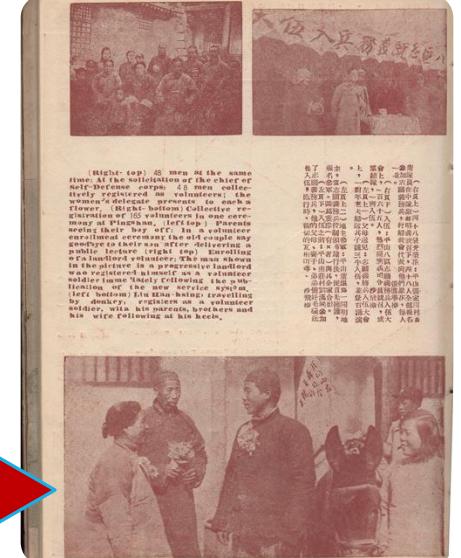


# Two Key Ways Computer Vision Compares and Contextualizes Historical Photographs

## 1. Juxtaposing Original And Printed Images

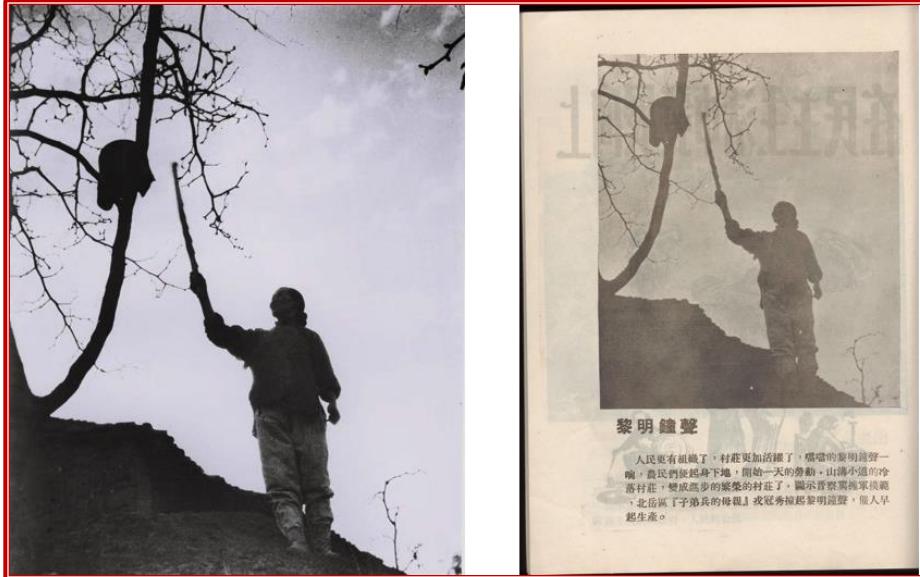


## 2. Tracking Photographic Circulation

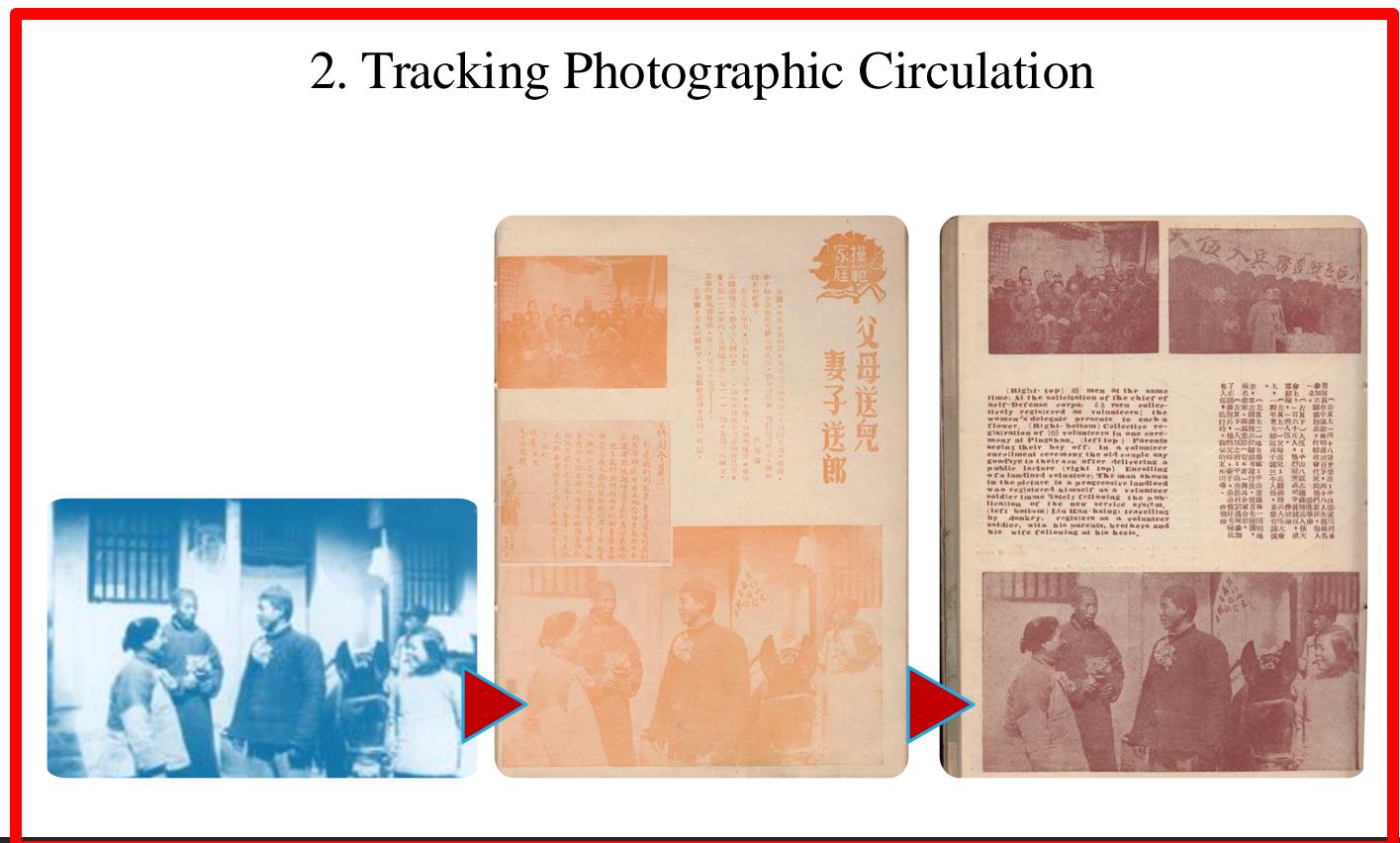


# Two Key Ways Computer Vision Compares and Contextualizes Historical Photographs

## 1. Juxtaposing Original And Printed Images

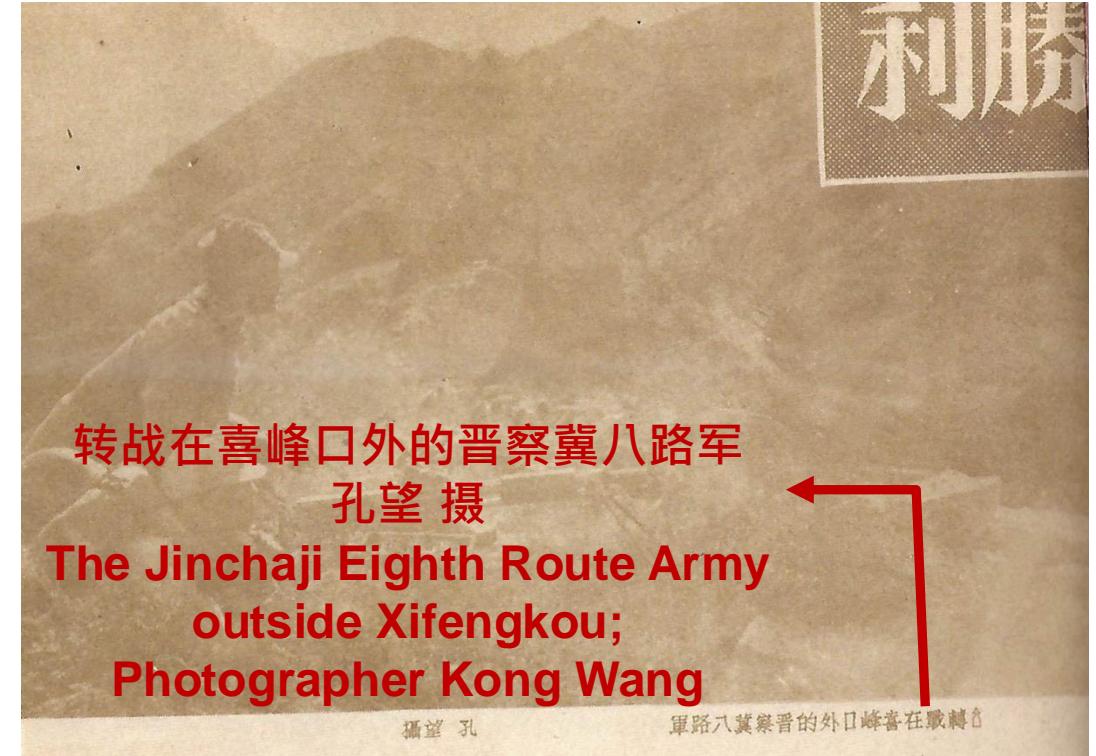


## 2. Tracking Photographic Circulation





## Case Study 1: Intentional Misinformation in Image Captioning



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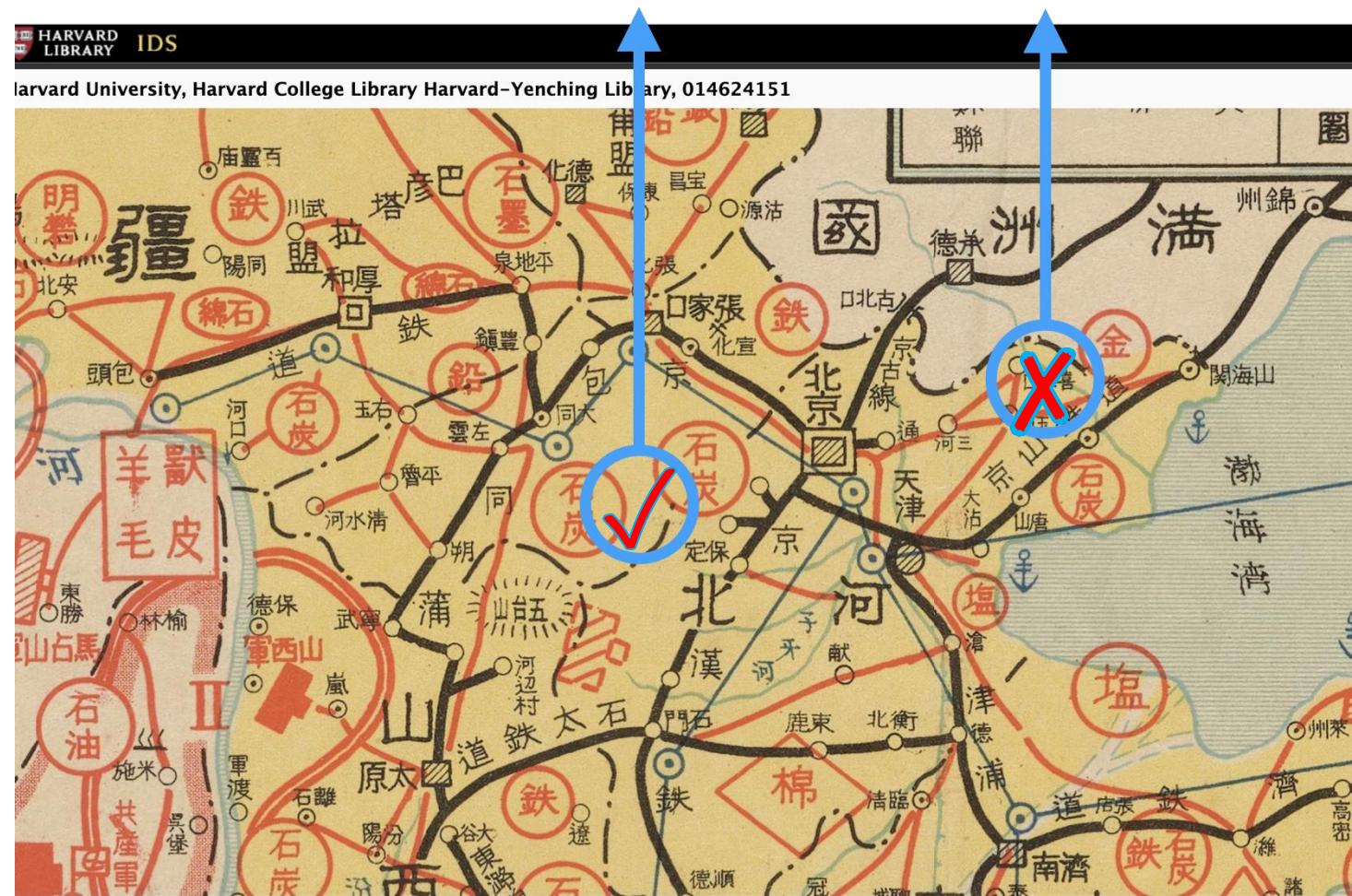
# Fact Check

Scholar Si Sushi [Si 2016] has confirmed that this photographs was actually taken by photographer Sha Fei and the real shooting location was in Futuyu, Laiyuan County, Hebei Province, rather than Xifengkou in Qianxi County, Tangshan, Hebei Province.

"The Anti-Japanese War Zone and Resource Transportation Network in China (支那抗日戰區及資源交通網要圖)," Yellow Region is Japanese-Army occupied area; This map was created by Japanese in 1941; Provided by Harvard University Library.

Futuyu in Laiyuan County, Hebei Province

Xifengkou in Qianxi County, Tangshan, Hebei Province



# Mao's 721 Policy

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A criticism against Mao for not fighting the Japanese army, a belief endorsed by the government of the Republic of China and the Kuomintang, as well as other groups opposing CCP.



**Development**



**Compromise**



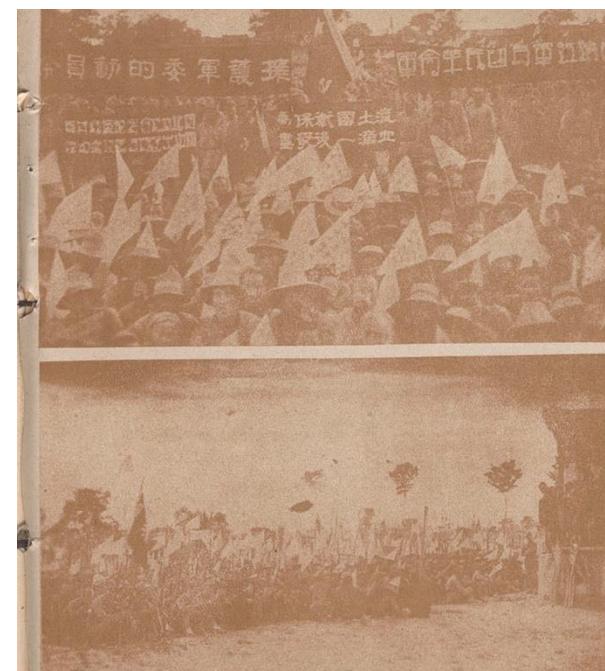
**Resistance**

# Mao's 721 Policy

## 7: Development

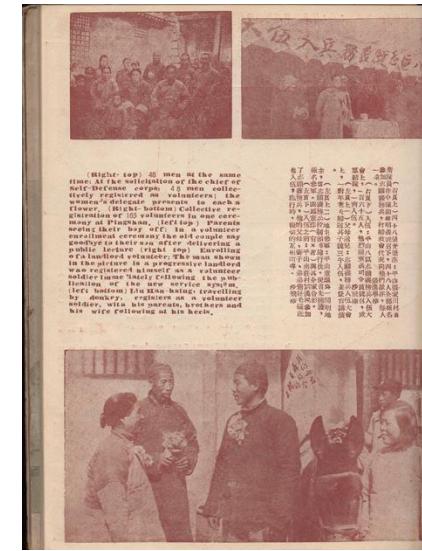
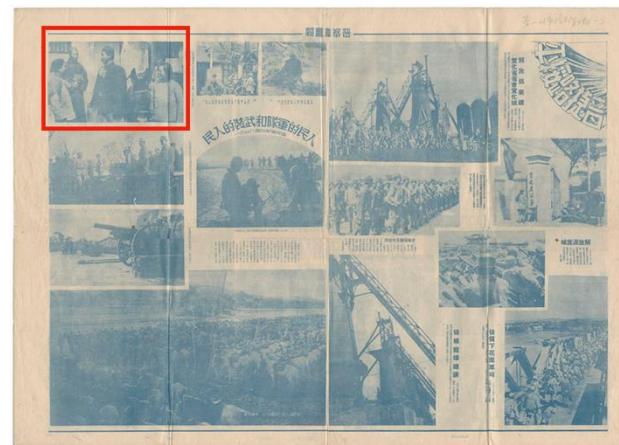
## 2: Compromise

## 1: Resistance





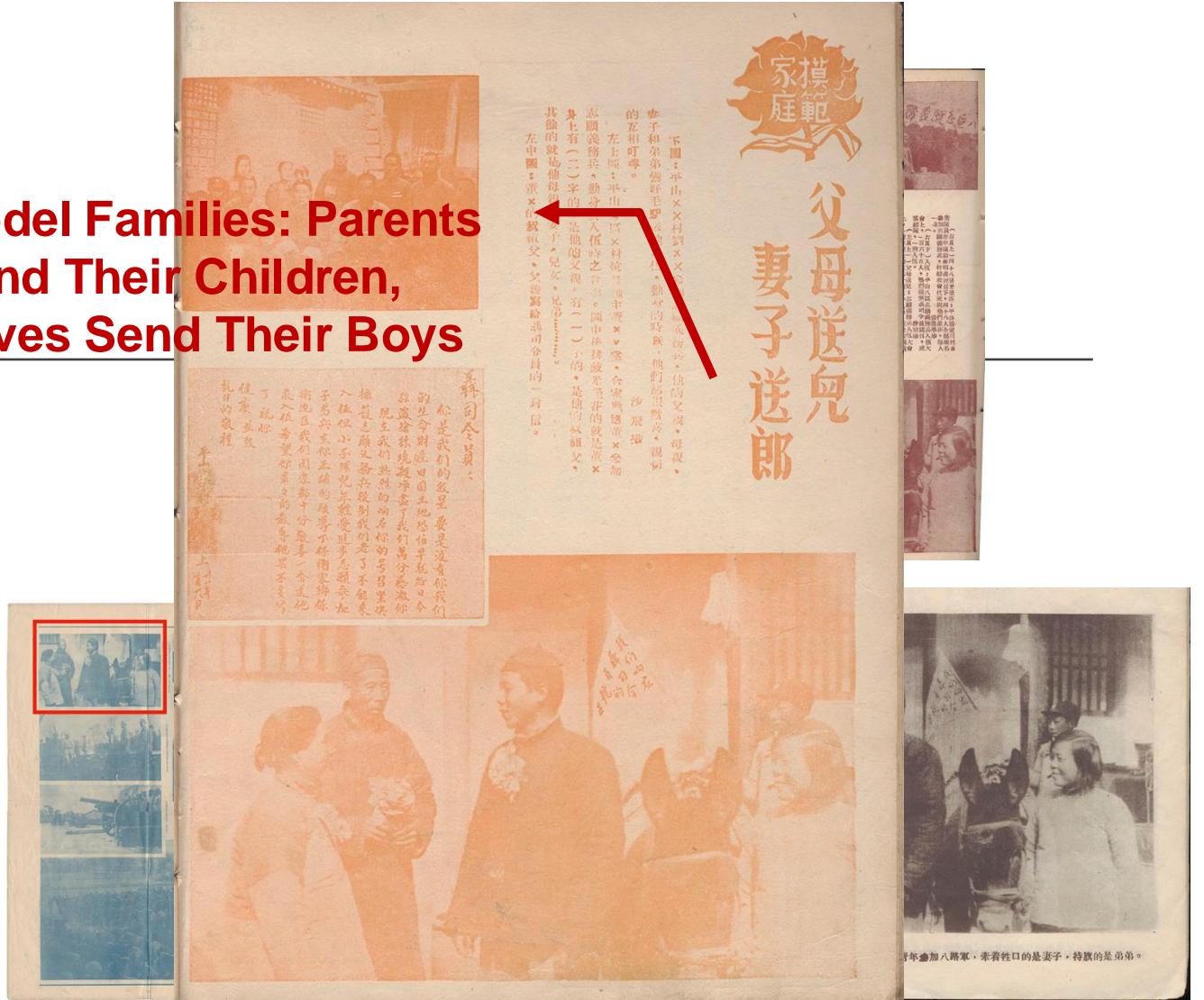
## Case Study 2: Resource Scarcities and Image Manipulation



## Case Study 3: Editorial Changes in Image Contextualization



## Model Families: Parents Send Their Children, Wives Send Their Boys



Case Study 3: Editorial Changes in Image Contextualization



# People of the Border Areas: Join the Volunteer Conscripts in Large Numbers

(Right-top) 48 men at the same time: At the solicitation of the chief of Self-Defense corps, 48 men collected themselves as volunteers; the government presents to each a flower. (Right-bottom) Collective registration of 165 volunteers in one ceremony at Pingshan. (Left-top) Parents send their boy off: In a volunteer army ceremony the old couple say goodbye to their son after delivering a public lecture (right-top) Enrolling of a landlord volunteer: The man shown in the picture is a progressive landlord who registered himself as a volunteer soldier immediately following the publication of the new service system. (Left-bottom) Liu Han-hsing, travelling by donkey, registers as a volunteer soldier, with his parents, brothers and his wife following at his heels.



## Case Study 3: Editorial Changes in Image Contextualization



## Case Study 3: Editorial Changes in Image Contextualization



## Case Study 3: Editorial Changes in Image Contextualization





# Summary

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- ❖ The importance of context in analyzing historiography and image editing
- ❖ The fluidity and adaptability of photographs as historemes: photographs and accompanying captions can be manipulated or adapted to various contexts within the realm of propaganda.
- ❖ Our computer vision pipeline can be used in combination with contextual analysis, a traditional media studies approach, to compare images and map the publication and circulation history of photographs.
- ❖ Our paper: Du, Lin, Brandon Le, and Edouardo Honig. “Probing Historical Image Contexts: Enhancing Visual Archive Retrieval through Computer Vision.” *ACM Journal on Computing and Cultural Heritage* 16, no. 4: 84:1-84:17. <https://doi.org/10.1145/3631129>

# The Power of Computer Vision in Historical Analysis

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Role of computer vision in identifying and tracking image modifications and circulations

Looking into the complex intersections of aesthetics, politics, and misinformation in image circulation

# Project 1: Detecting Published vs. Unpublished Images in Wartime Archives





# Project 1: Goal

- Use supervised learning to classify whether historical photographs from the Manchuria Railway archive were published in North China Magazine or remained unpublished.
- Why?
  - Many wartime photos were taken, but only some were published.
  - By identifying which were published or even repeatedly published, we can study how visual culture was shaped by editorial choices, aesthetic preferences, and political agendas.
- This is part of a broader project on AI-assisted visual historiography — and we may turn this into a publishable research paper together.

# Dataset

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- Kyoto University Archive: ~40,000 wartime photographs (1930s–1940s)
- North China Magazine: Digitized magazines from the same era
- Preprocessed Matches: Your instructor's pipeline already matched each archive photo to its "top 10 most similar" magazine images using self-supervised learning (no human labels needed for matching).

# Task

- Problem: The current pipeline has false positives (incorrect matches).
- Your Job: Improve accuracy by filtering bad matches using Aspanformer and supervised learning.
- How?
- Step 1: Label Training Data:
  - If a photo has  $\geq 1$  correct match in the magazine  $\rightarrow$  label as published.
  - If all matches are incorrect  $\rightarrow$  label as unpublished.
  - This becomes your labeled dataset (I already have 66 labeled matches, but we can label more if necessary).

# Task

- Step 2: Deploy Aspanformer to calculate local feature matches (keypoints) between archive photos and magazine images.
  - Install Aspanformer: Follow instructions at <https://aspanformer.github.io/>.
  - Save Results: For each archive photo, record the maximum number of matching points among its top 10 matches.
    - Matching points reflect how much two images overlap in local features.
    - Published photos (even if cropped/edited) will share many keypoints with their magazine versions.
    - Unpublished photos will have few/no matching keypoints.

# Task

- Step 3: Determine the Threshold
  - Correct matches will likely have high matching points (e.g.,  $>50$ ).
  - False positives will have low matching points (e.g.,  $<20$ ).
  - Set a threshold (e.g., 40 matches) to separate the correct matches.

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Image

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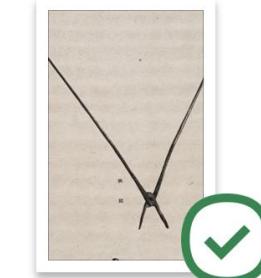
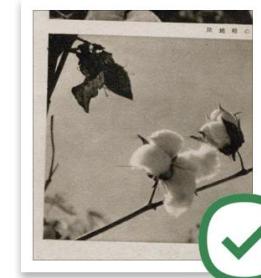
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Downloads

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# Reference

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硕鼠

爾雅鼯鼠郭璞註形大如鼠頭似兔尾  
好在田中食粟豆關西呼爲鼴鼠見廣

上屋  
能穴

相鼠



鼫鼠

文子聖人師拱鼠制禮錦馬記拱鼠行  
則拱手而立捕之卽跳躍走去愚按碩  
爾雅鼯鼠也



名物圖說

角弓

卷二 獸

毛傳猱猿屬鄭箋猱之性善登木正義

輩屬非猱也陸璣云猱猶猴也楚人謂  
爲玃長臂者爲玃玃之性腰者爲獮玃

於獮猴然則猱玃其類同埤雅玃玃  
小類猿長尾尾作金色俗謂之金冶  
山中人以藥矢射之取其尾爲臤禊  
愛其尾中矢毒卽自齧斷其尾以擲

頃氏以爲

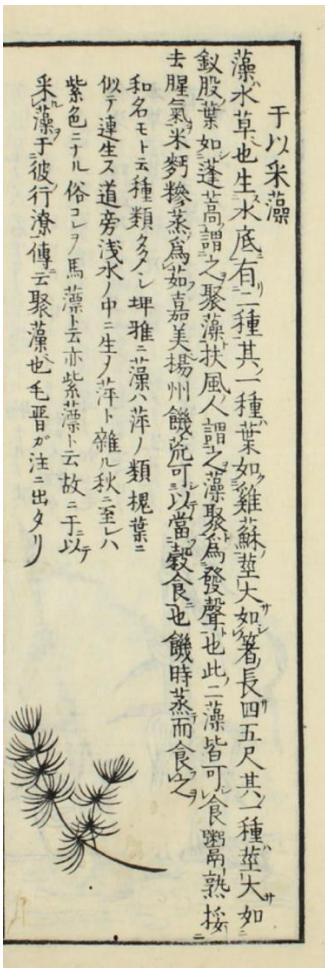
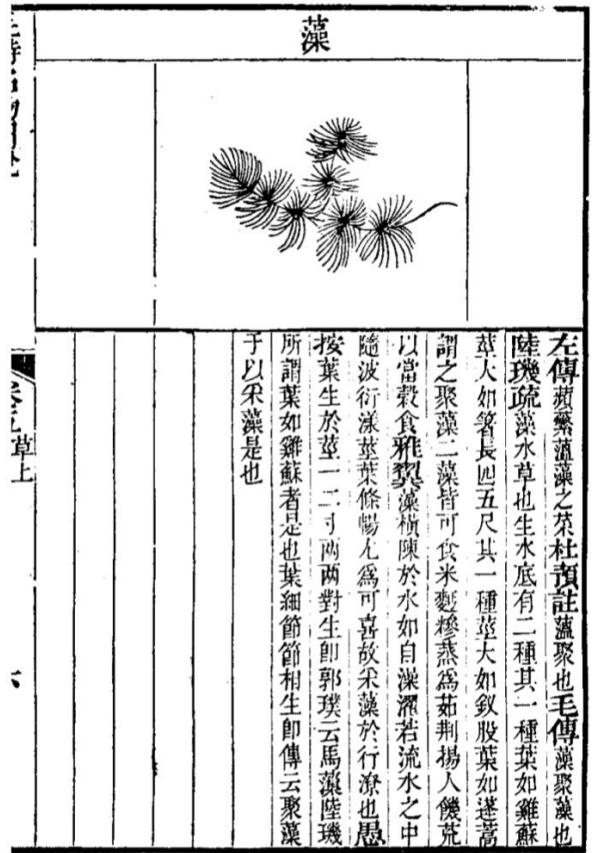
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Project 2: Similarities Between Illustrations

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# Goal

- Explore how Book of Songs (Shijing 詩經) plant and animal imagery was visualized and interpreted differently in Qing China and Tokugawa-to-Meiji Japan through annotated illustrations.
- When Chinese and Japanese commentaries both include illustrations for the same object (e.g. 荀菜 xìng cài, 蘋 píng),
  - → Do they visualize the object in a similar way?
- Did Japanese illustrators copy Chinese visualizations? Or did they reinterpret them in unique ways?
- We may turn this into a publishable research paper together too.

# Tasks

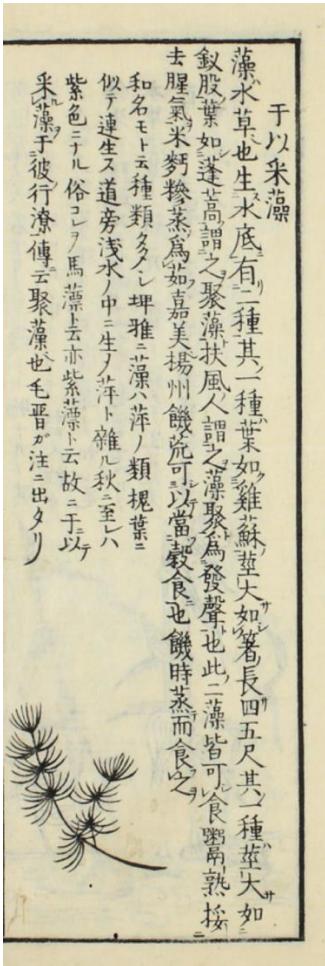
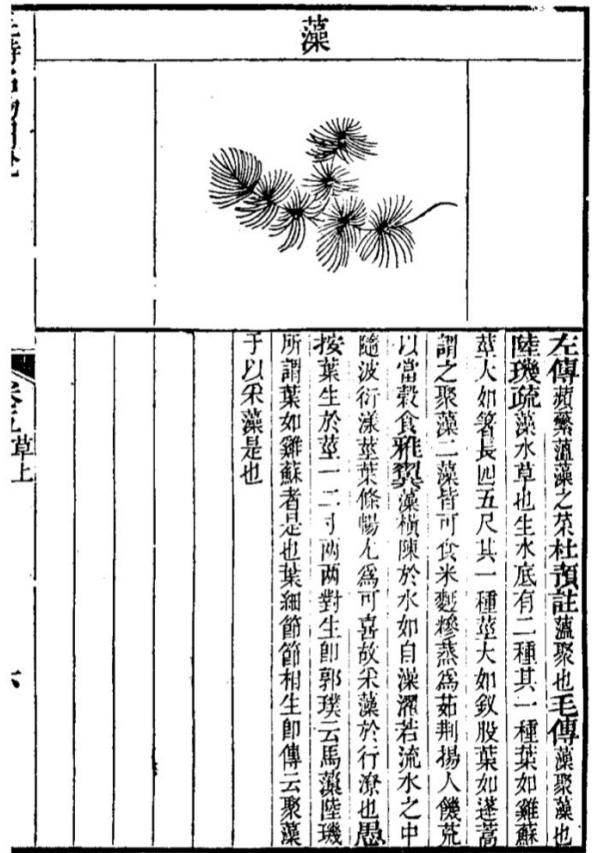
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- Goal: Use unsupervised learning to analyze how the same plant/animal from the Book of Songs (Shijing) was illustrated differently in Chinese (Qing Dynasty) and Japanese (Tokugawa-to-Meiji era) commentaries.
- Dataset:
  - Chinese: 毛诗名物图说 (Qing Dynasty, Xu Ding).
  - Japanese: 毛诗品物图考 (Okamoto Ryūho) and 陸氏草木鳥獸虫魚疏圖解.

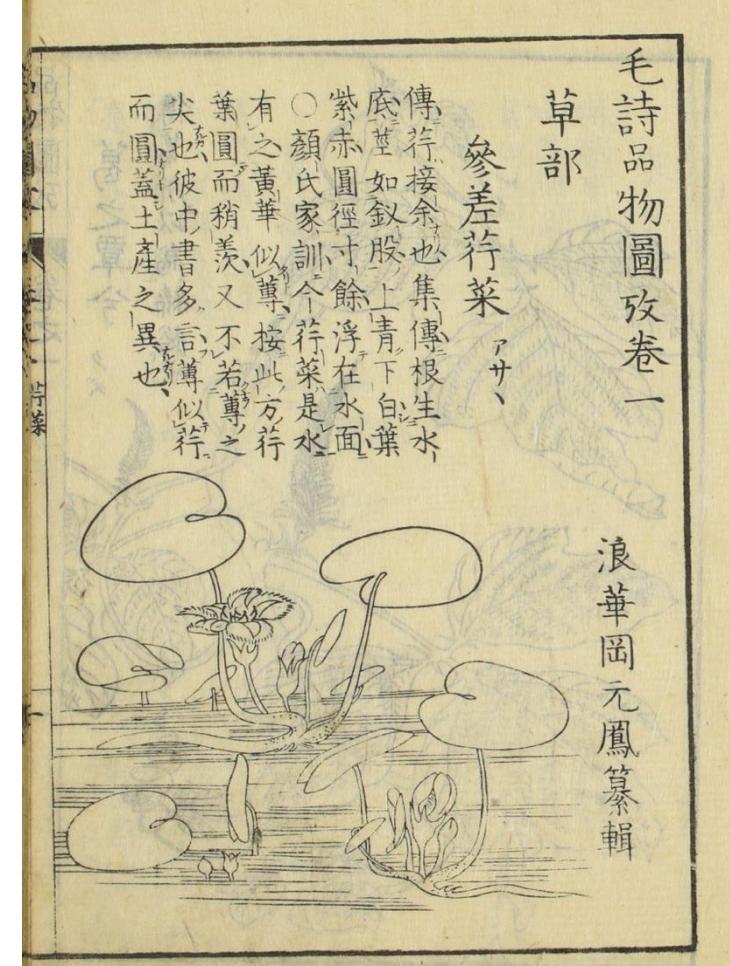
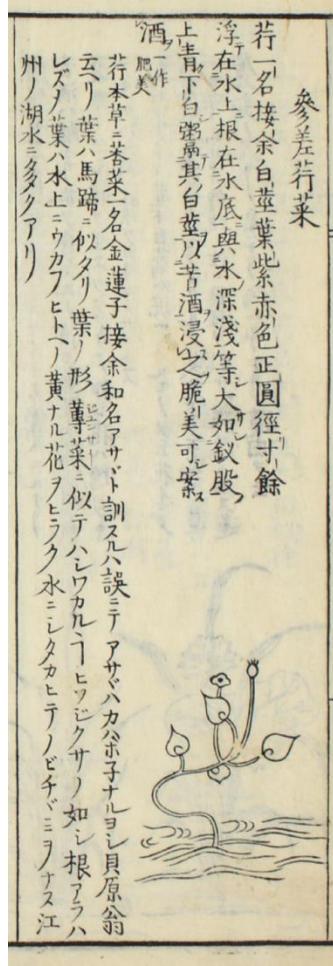
# Task

- Step 1: Data Preparation: Manually crop illustrations from scanned book pages
- Step 2: Visual Comparison
  - 1. Extract Image Features: Use a pretrained CNN (e.g., ResNet-18) to convert images into numerical vectors (embeddings).
    - Alternative: Use traditional features such as SIFT for simpler illustrations.
  - 2. Calculate Similarity Scores: For each Chinese-Japanese pair, compute cosine similarity between their embeddings.
    - High score = Similar visuals; Low score = Different visuals.
  - 3. Rank Pairs: Sort all object pairs by similarity scores (most to least similar).

# 藻



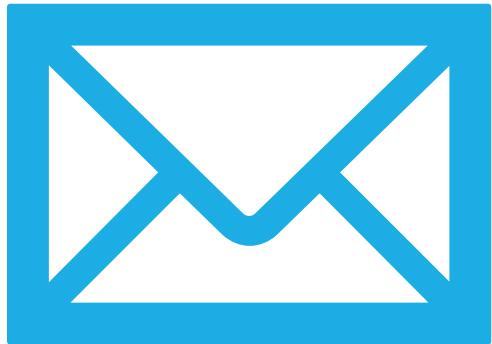
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# Reference

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If you're interested in working with either project, please contact me at [dulin525@gmail.com](mailto:dulin525@gmail.com)/[WeChat](#): dulinlindu. I can share the datasets with you and provide Google Colab Pro Account.

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## CONCLUSION: DIGITAL HISTORICAL FORENSICS

- Digital Historical Forensics, particularly through AI, bridge different visual media including pictorials and exhibitions with textual media to track media circulation and enhance media studies.
- Its importance in Japan and China studies lies in discovering how Chinese photojournalists and editors, despite limited material resources, effectively made history and competed with Japanese propaganda efforts.
- The methodology is generalizable and extendable across diverse cultural contexts and archives, facilitating cross-regional research.
- This approach enables robust source criticism at a new scale, applicable to both historical and modern media, and encourages the digitization of offline archives.
- Contact: [dulin525@gmail.com](mailto:dulin525@gmail.com)