

# Deterministic Routing between Layout Abstractions for Multi-Scale Classification of Visually Rich Documents

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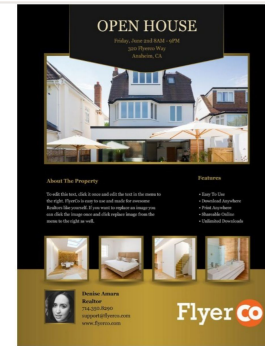
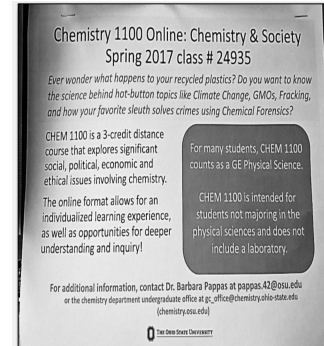
# Outline

- ❖ Visually Rich Documents
- ❖ Challenges
- ❖ Our contributions
  - Spatial Pyramid Model
  - Deterministic Routing Scheme
- ❖ End-to-end architecture for classification
- ❖ Experimental results
  - Tobacco-Litigation dataset
  - Outperforms previous state-of-the-art by 4.73%



# Visually Rich Document

- Visual cues to highlight distinct semantic entities
  - Color
  - Font size
  - Text orientation and positioning
  - Symmetry etc.
- Classification is a precursor of many document understanding tasks
  - Indexing
  - Information extraction



# Our objective

1. Maximize classification accuracy using visual (dis)similarity in layouts
  - Multi-scale features
    - Encode local invariant patterns from multiple resolutions of the document
    - More robust than single-scale features: Leads to better classification accuracy
2. Minimize end-to-end inference turnaround
  - Suitable for modern interactive workflows

Assumption: A soft constraint on the relative positioning of components in documents belonging to the same class



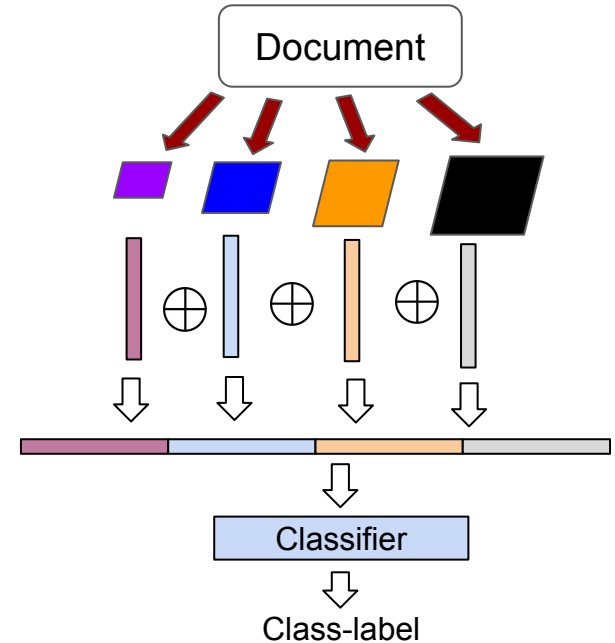
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# Challenges

- **Multi-scale: Aggregation based methods**
  - Increased overhead in inference turnaround
  - Marginal gain in accuracy from single-scale counterparts
- **Infeasible when bound by near real-time latency**
  - e.g. Interactive workflow on edge-devices



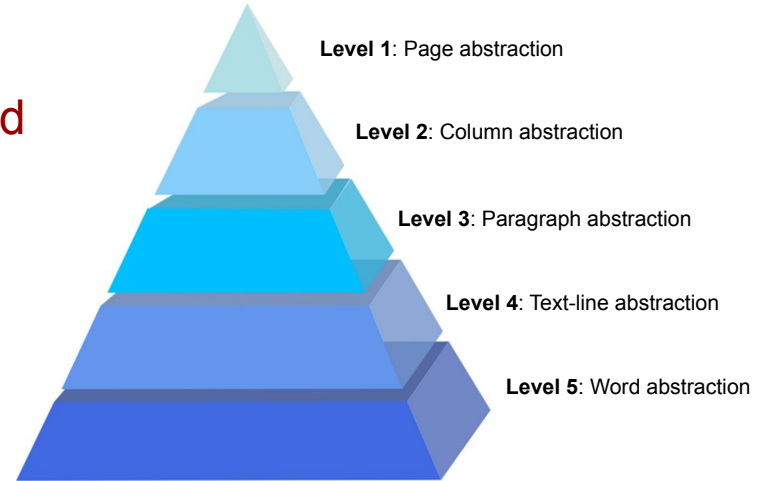
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# A Spatial Pyramid Model for document representation

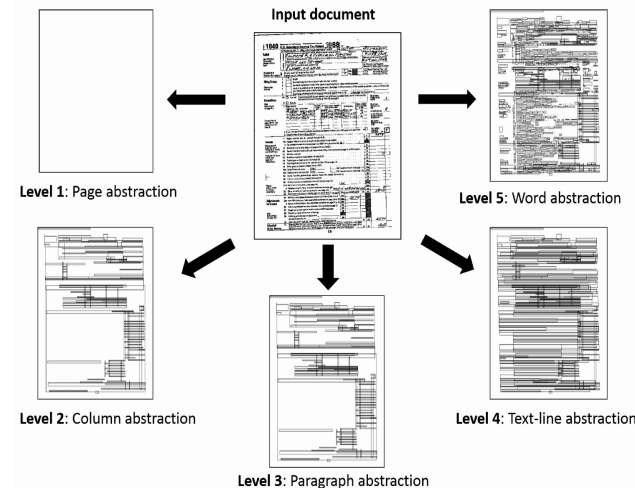
- Objective: Increase accuracy by extracting highly discriminative features
- Represent the document as a spatial pyramid
  - Each level corresponds to an abstraction in layout hierarchy:
    - Words
    - Text-line
    - Paragraph
    - Column
    - Page





# Layout abstraction at each level of the Spatial Pyramid Model

- Each pyramid level corresponds to an abstraction at that level of the layout hierarchy
  - Binary Image with same dimensions as the input document
  - e.g. Word abstraction
    - Bounding box corresponding to each word of the document
- Layout hierarchy derived using an open-source page segmentation algorithm <sup>1,2</sup>



[1] <https://github.com/tesseract-ocr/tesseract>

[2] Smith, Ray. "An overview of the Tesseract OCR engine." In *Ninth International Conference on Document Analysis and Recognition (ICDAR 2007)*, vol. 2, pp. 629-633. IEEE, 2007.

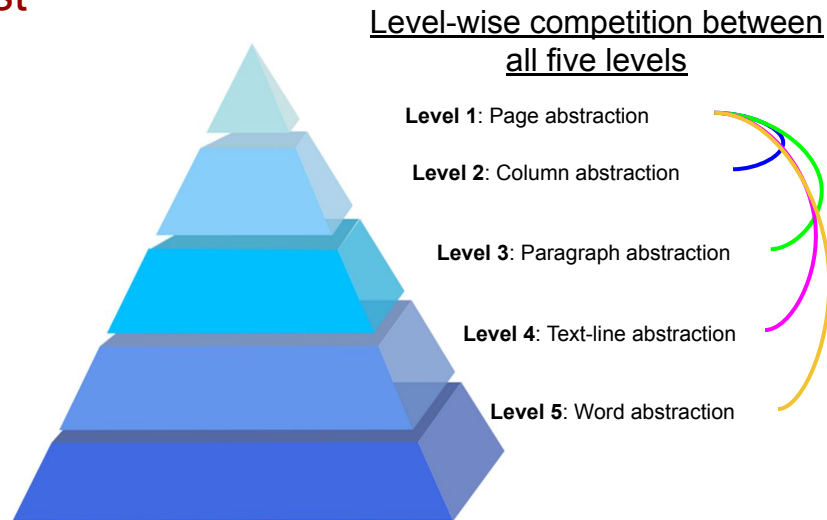
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# A Deterministic Routing Scheme for feature selection

- Attention-like operation to select the most discriminative pyramid level
- Introduce level-wise competition
  - Select the most discriminating pyramid level
- The winner represents the document for final prediction



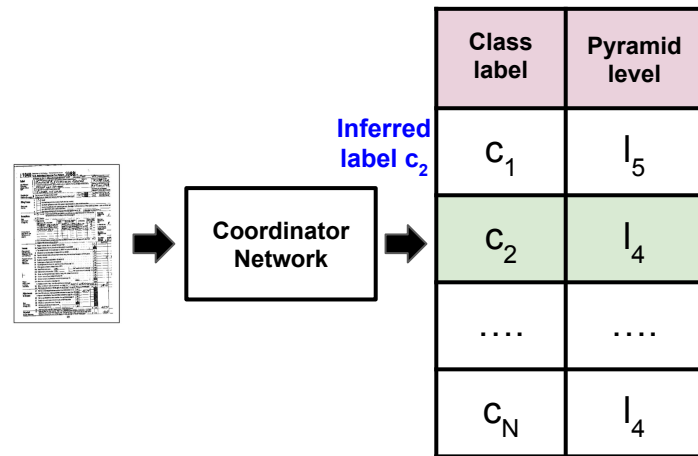
# Supervised coordinator network for Deterministic Routing

- **LadderNet: Softmax classifier network**

- Extended the MobileNetV2 architecture <sup>1,2</sup>
- Fast: Factorized convolution
- Pretrained on ImageNet, fine-tuned on training corpus

- **Softmax label to query the multiplex table**

- Single entry for each class-label
- Route to the corresponding pyramid level



[1] Howard, Andrew G., Menglong Zhu, Bo Chen, Dmitry Kalenichenko, Weijun Wang, Tobias Weyand, Marco Andreetto, and Hartwig Adam. "Mobilenets: Efficient convolutional neural networks for mobile vision applications." *arXiv preprint arXiv:1704.04861* (2017).

[2] Sandler, Mark, Andrew Howard, Menglong Zhu, Andrey Zhmoginov, and Liang-Chieh Chen. "Mobilenetv2: Inverted residuals and linear bottlenecks." In *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*, pp. 4510-4520. 2018.

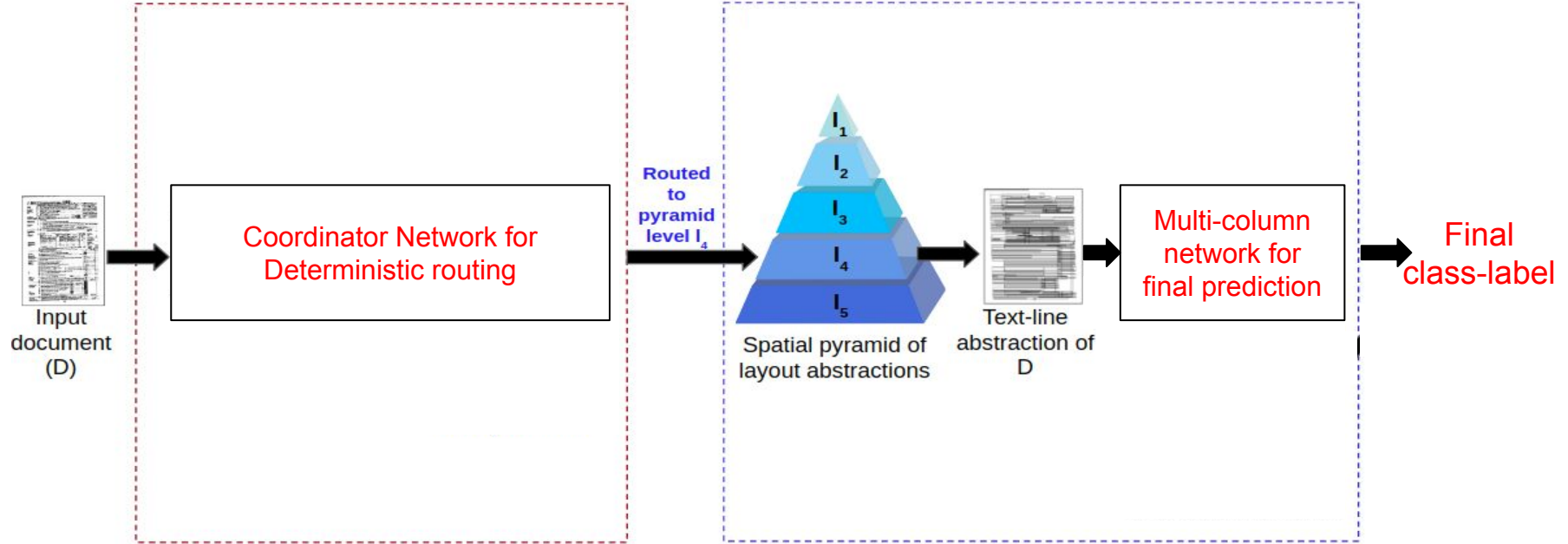


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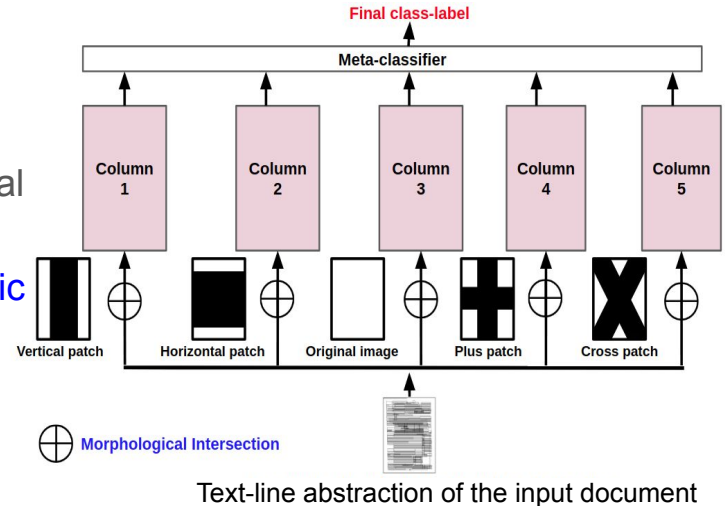
# End-to-end classification workflow



# A Multi-column Network for Final prediction

- **Multi-column neural architecture**

- Depth-separable convolutional columns
- Input: Morphological transformation of the document
- Combine predictions from each column to compute final class label
  - **Meta-classifier**: Simple-avg, weighted-avg, logistic regression and MLP



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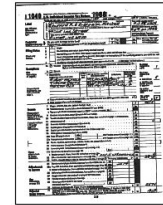
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# Experiments

- Four publicly available datasets of single-page documents
  - NIST Special dataset-6 (easy)
  - Medical Article Records Groundtruth or MARG dataset (medium)
  - Tobacco Litigation dataset (hard)
  - IIT-CDIP dataset (hard)
- Evaluation metrics
  - Classification accuracy
  - Inference turnaround time



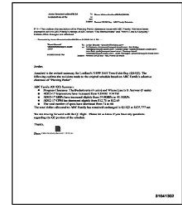
(A1) NIST



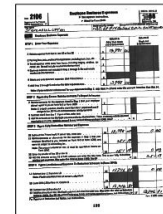
(B1) MARG



(C1) Tobacco



(D1) CDIP



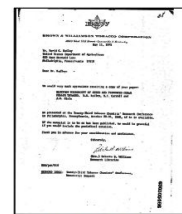
(A2) NIST



(B2) MARG



(C2) Tobacco



(D2) CDIP

# Result highlights

- Tobacco Litigation dataset
  - High inter-class and low intra-class layout similarity
  - 3482 documents, 10 document classes (e.g. 'letter', 'memo', 'email' etc.)
- Median classification accuracy over 25 trials := 82.78%
  - Absolute improvement over previous state-of-the-art := 4.73%
- Average inference turnaround time := 362 ( $\pm 10.27$ ) ms
  - Average speed-up factor > 6

More results and ablation study in paper !!



# Takeaways

- Multi-scale classifier for visually rich documents
  - Spatial pyramid model
    - Highly discriminative features → Increase in accuracy
  - Level-wise competition for optimal document representation
    - Attending to the most discriminative layout abstraction improves accuracy
- Inference turnaround time < interactive latency ( $\approx 500$  ms)
- Robust performance on four publicly available datasets



Thank you!