Project presentation

Image analysis and pattern recognition

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    iii. compute location of the arrow in each frame, defined as the center of the red mask
    iv. store consecutive locations in list `arrow_locations`:

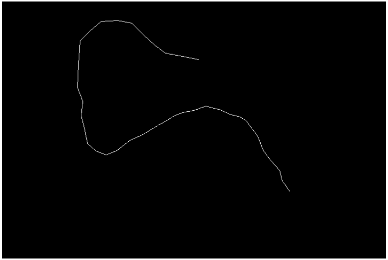

Figure: Arrow path

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3. with the obtained windows, create list of dictionaries that have two keys: `image_box` and `center` (mean of white pixels)

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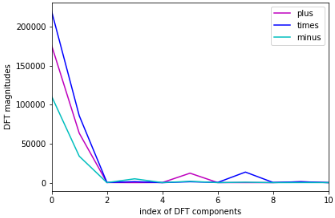

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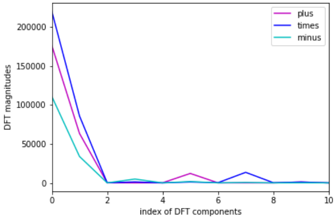

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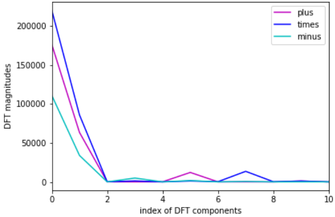

- `classify_number`: CNN trained on augmented MNIST
  - problems to tackle: digits are hand-written, have varying contours, can be rotated, scaled, have a non-centered sliding window

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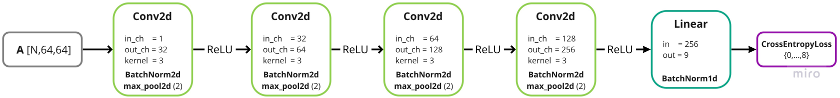

Figure: Architecture of the CNN

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   2. at each iteration, write current state of formula