

Deep Learning for Automatic Localization, Identification, and Segmentation of Vertebral Bodies in Volumetric MR Images

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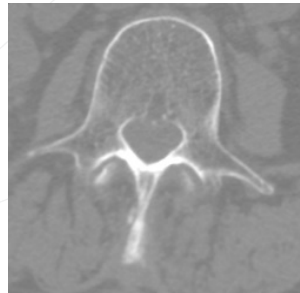
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Purpose

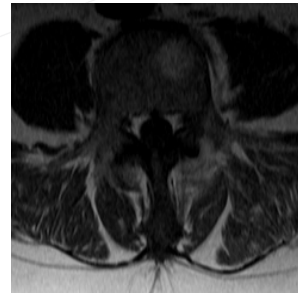
- Localization and identification of vertebrae.
 - Core requirement for computer-aided systems for spine.
- Segmentation of vertebral structures enables quantitative analysis of spine pathologies.
 - deformations caused by different pathologies

Challenges

- Repetitive nature of vertebral column.
- Variation in terms of field of view.
- Poor contrast of bone structures in MR images.
- Variation in surrounding soft tissue contrast in MR images.
- Magnetic field inhomogeneity in MR images.
- Large inter-slice gap (around 4 mm) in typical clinical MR images compared to CT.



CT



MR

Related Work in Localization

- Mostly make assumptions about which vertebrae are visible in the scan.
 - [Ma'13], [Oktay'13], [Neubert'12], [Stern'11], [Naegel'07], [Schmidt'07].
- General methods are mostly developed for CT.
 - [Glocker'13], [Rasouliaan'13], [Glocker'12], [Klinder'09].
- Mostly are not integrated with segmentation.
 - [Glocker'13], [Glocker'12].

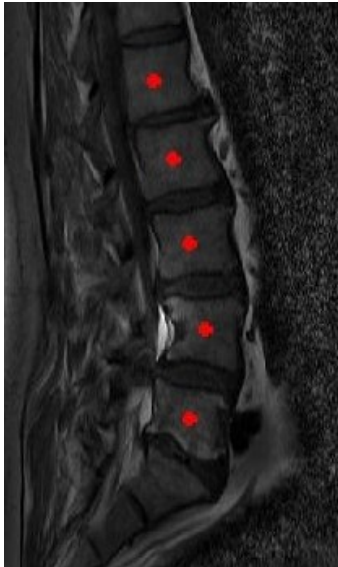
Related Work in Segmentation

- Mostly require user interaction.
 - [Suzani'14], [Rasouliau'13], [Hoad'02].
- Most approaches are in 2D
 - [Egger'12], [Carballido-Gamio'04], [Shi'07], [Huang'09].
- 3D Methods are mostly evaluated on MR images with inter-slice gap of 1.2 mm or less.
 - [Kadoury'13], [Neubert'12], [Stern'11].
- Each vertebra is mostly segmented independently.
 - [Neubert'12], [Stern'11], [Hoad'02].

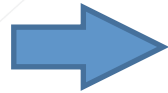
This Work

- We propose an automatic method for simultaneous localization and identification of vertebrae.
- The predictions are used for automating a registration-based segmentation technique.
- No assumptions are made about the visibility of specific vertebrae.
- Multiple vertebrae are registered simultaneously.

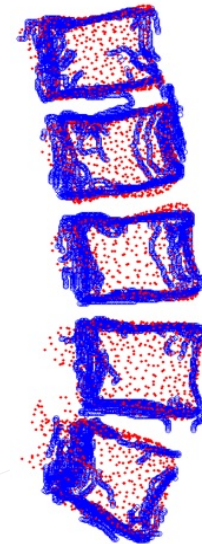
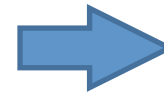
Method



**Automatic
Localization**



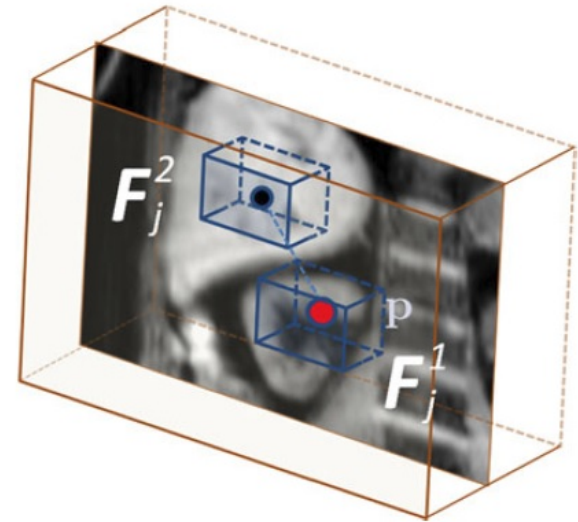
**Edge
Detection**



**Model
Registration**

Features

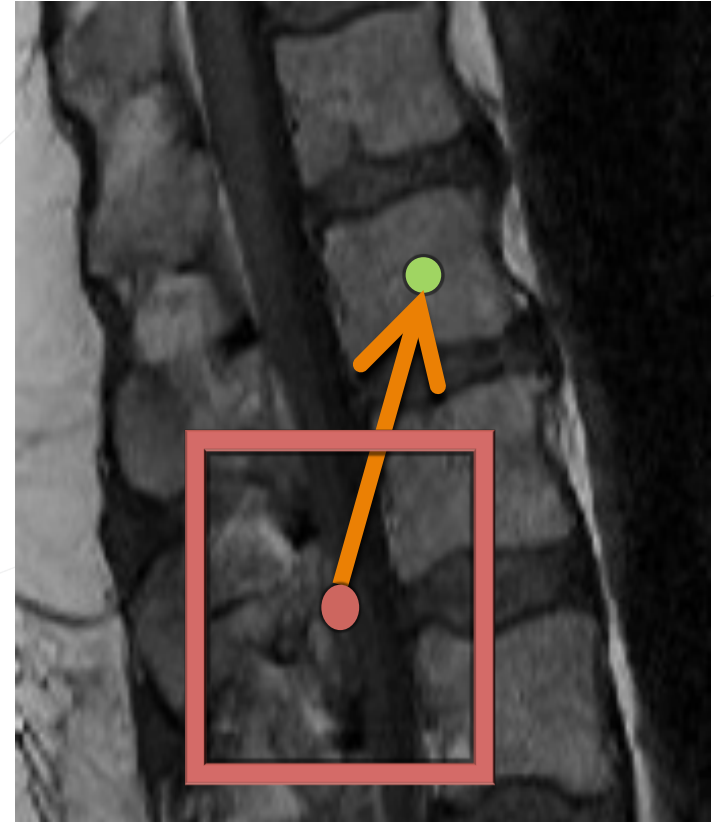
- 500 features are extracted for each voxel.
- Each feature: difference between the mean intensity over two cuboids displaced with respect to the reference voxel position.
- Feature dimensions and displacement are chosen randomly.



Intensity-based features

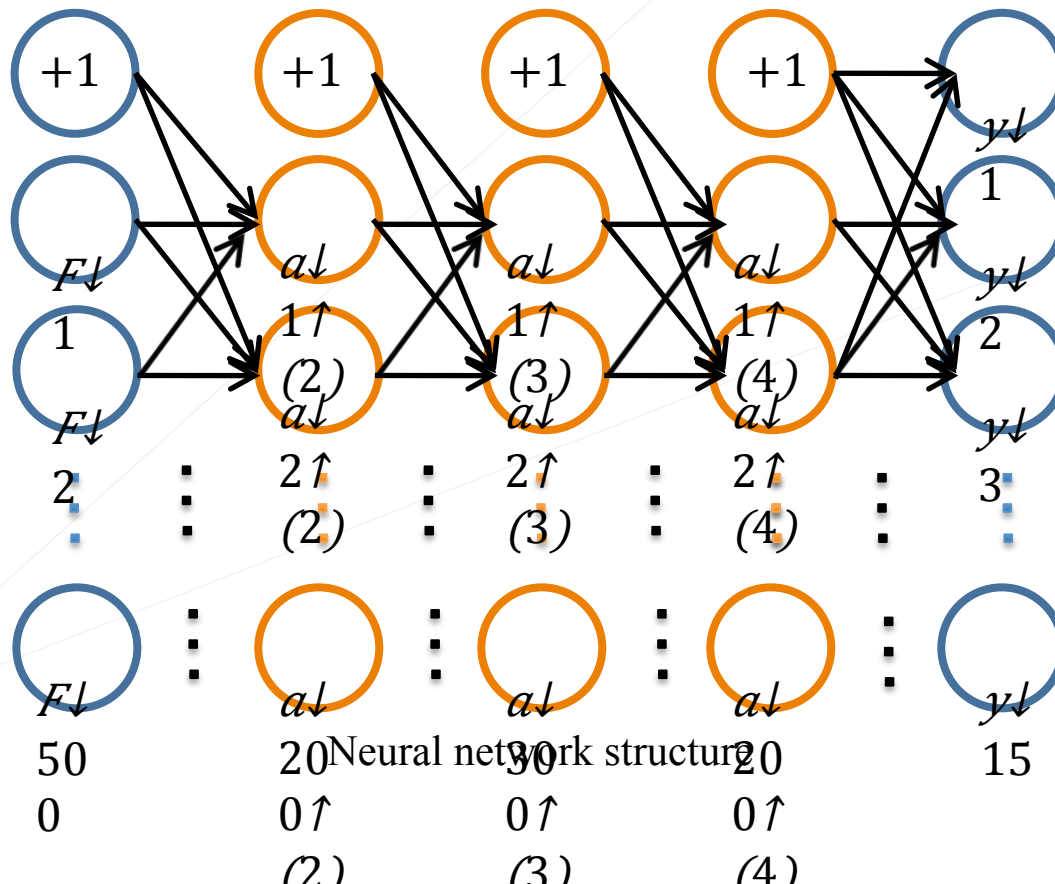
Problem Parameterization

- Vertebrae anatomy localization is parameterized as a multivariate regression problem.
- Each voxel votes for its relative distance to the centroid of each vertebral body.



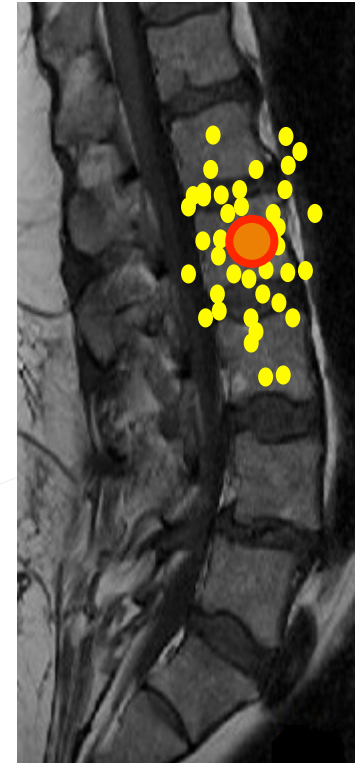
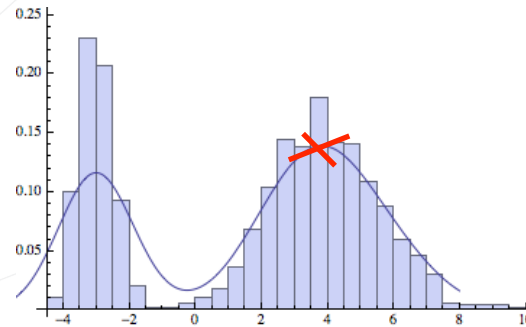
Deep Neural Network

- Neural network is trained using stochastic gradient descent and layerwise pre-training.



Centroid Estimation

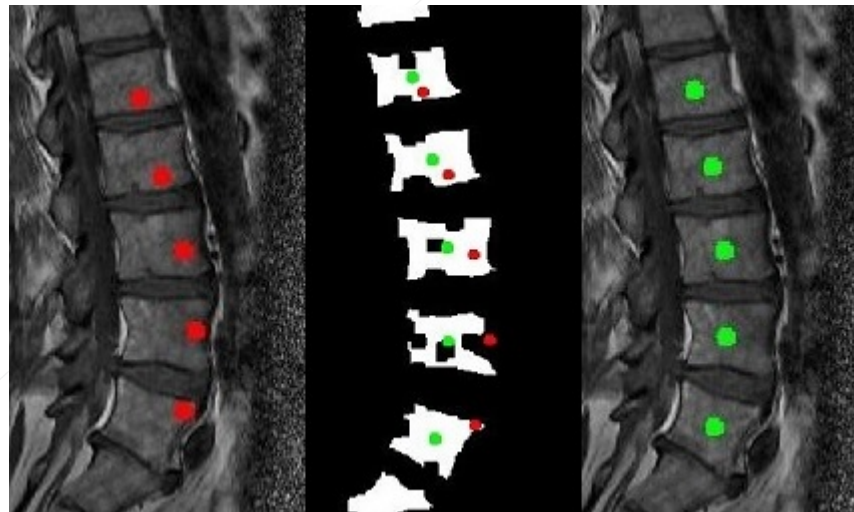
- Kernel Density Estimation is used for aggregating the votes of all voxels.



Centroid estimation

Refinement

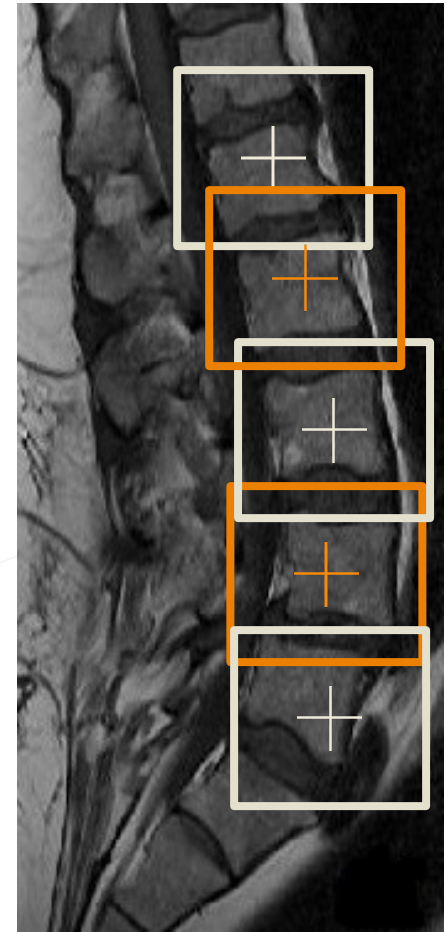
- Predicted points are refined by local Otsu thresholding. The points are replaced with the center of the closest large component.



Refinement step

Region of Interest

- Pre-processing steps and edge detection is only applied on boxes centered to the predicted points.



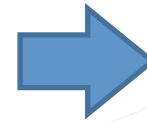
Predicted points and boxes

Pre-processing for Segmentation

- Intensity correction



Original image



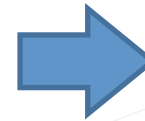
Intensity-corrected image

Pre-processing for Segmentation

- Anisotropic diffusion



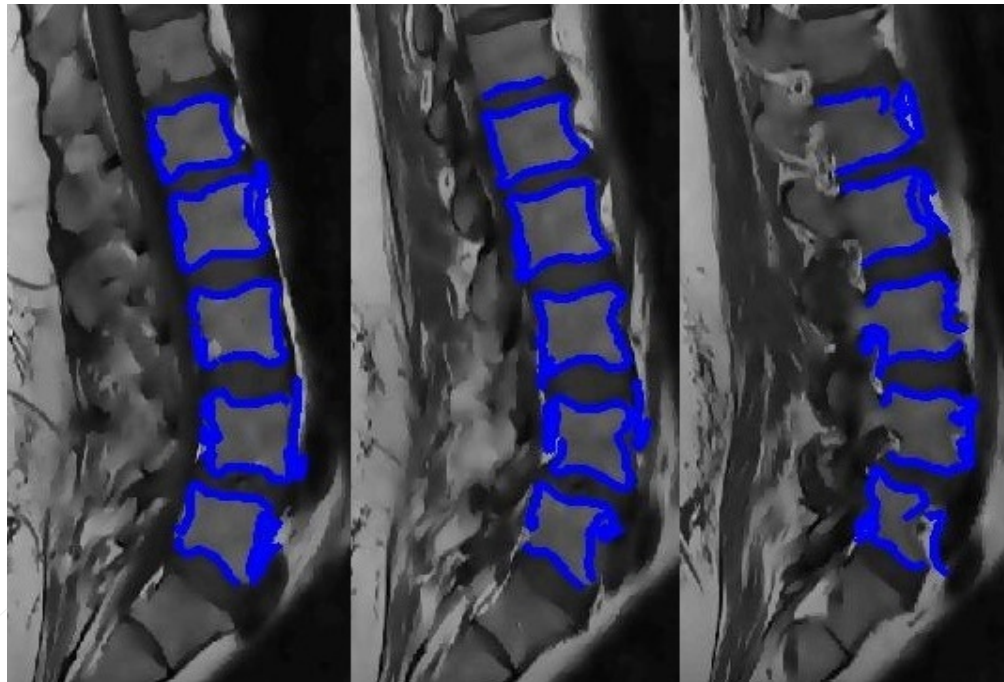
Intensity-corrected image



After anisotropic diffusion

Canny Edge Detection

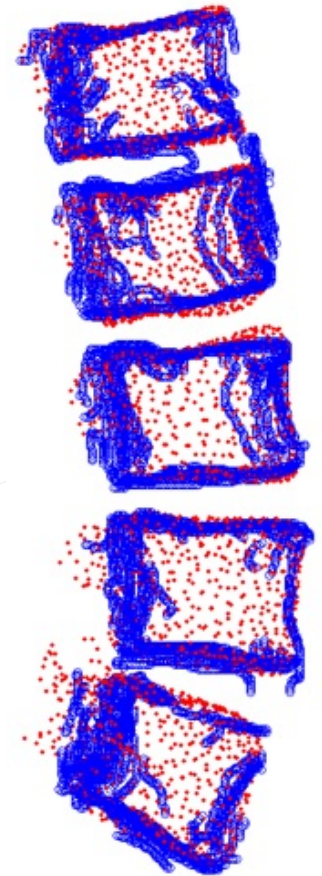
- Extract edges in the area of the predicted vertebrae.



Extracted edges using Canny edge detection on three slices of the same volume

Registration

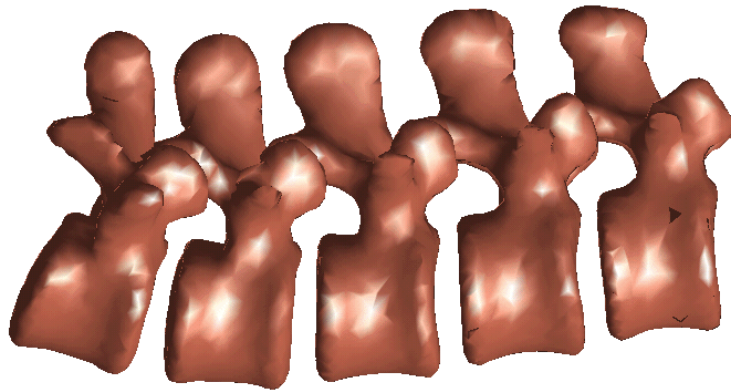
- Multi-vertebrae anatomical model is initialized by the predicted points from localization step.
- The model is registered to the edge map using an iterative expectation maximization method.
- Only vertebral body part of the model is used for registration.
- Correlation between shapes and poses of different vertebrae are taken into account.



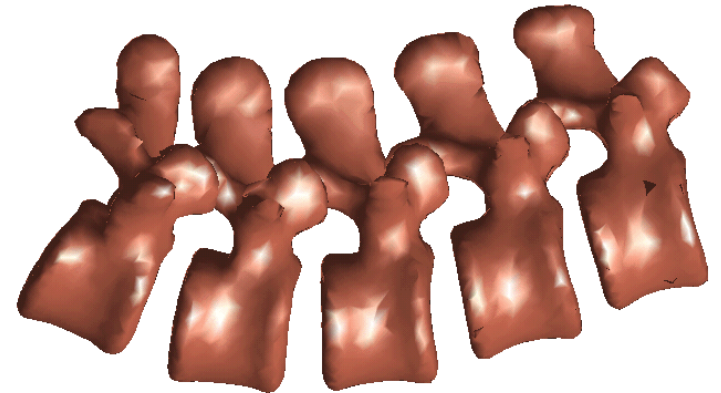
Registered model

Model Pose Variations

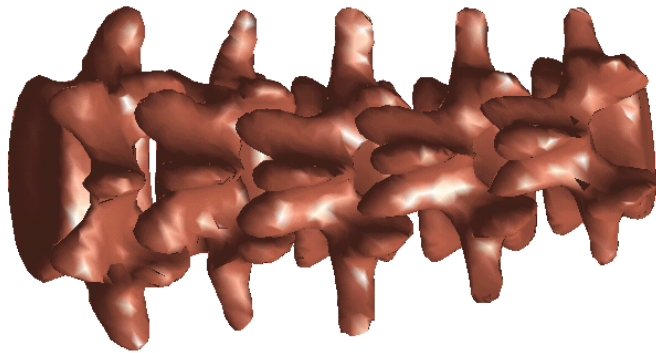
First Mode



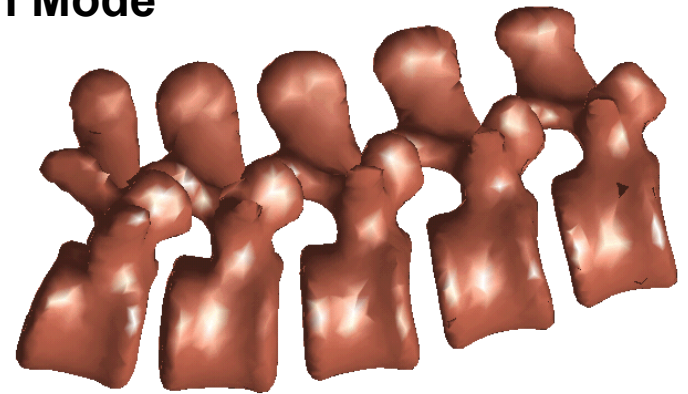
Second Mode



Third Mode

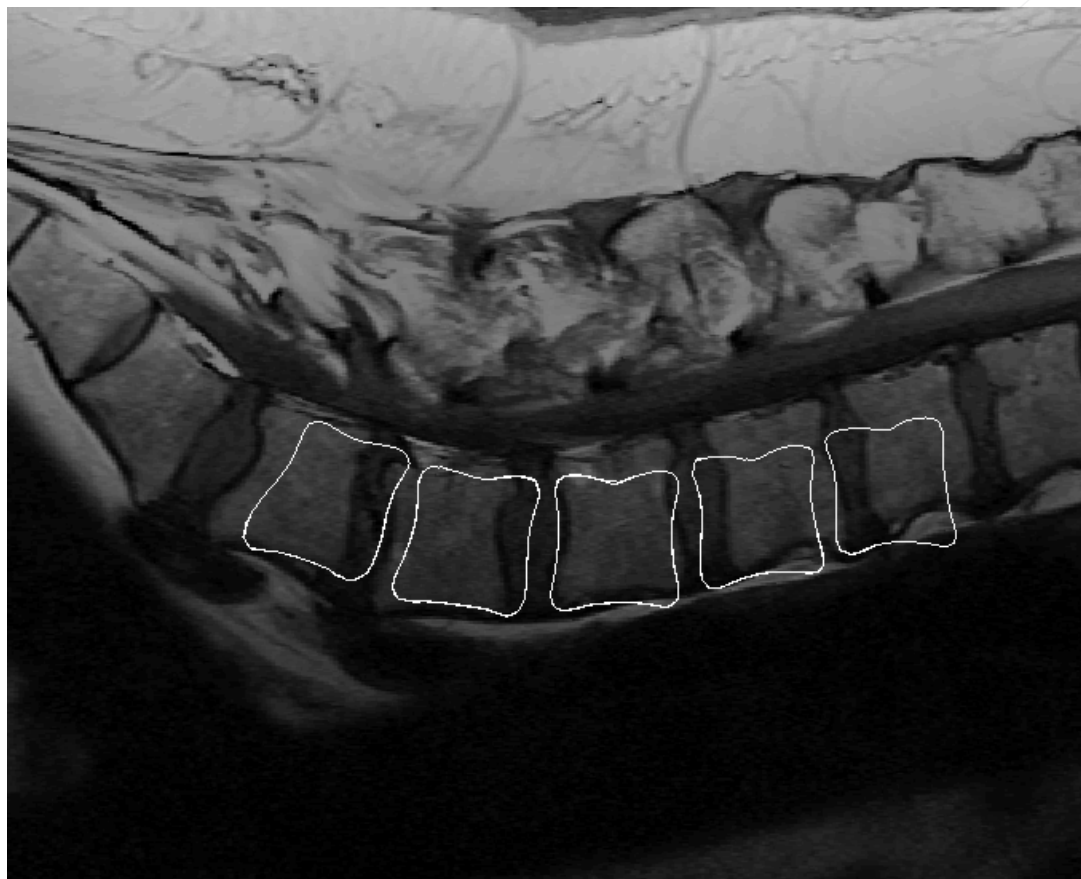


Fourth Mode



N=32

Registration



Registration on mid-sagittal slice

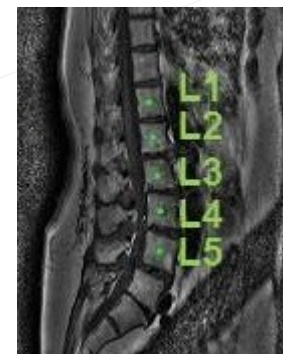
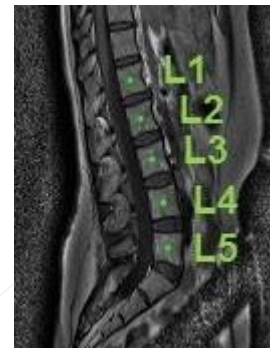
Results

- Evaluated on nine multi-slice MR images of lumbar spine.
- Inter-slice gaps in range of [3.3 mm 4.4 mm].
- Computation time: less than 3 minutes in total (on a 2.5 GHz Intel core i5 machine).

	Mean Error	Std	Identification
Deep learning localization	11.9 mm	6.3 mm	91 %
After refinement	3.0 mm	2.4 mm	100 %

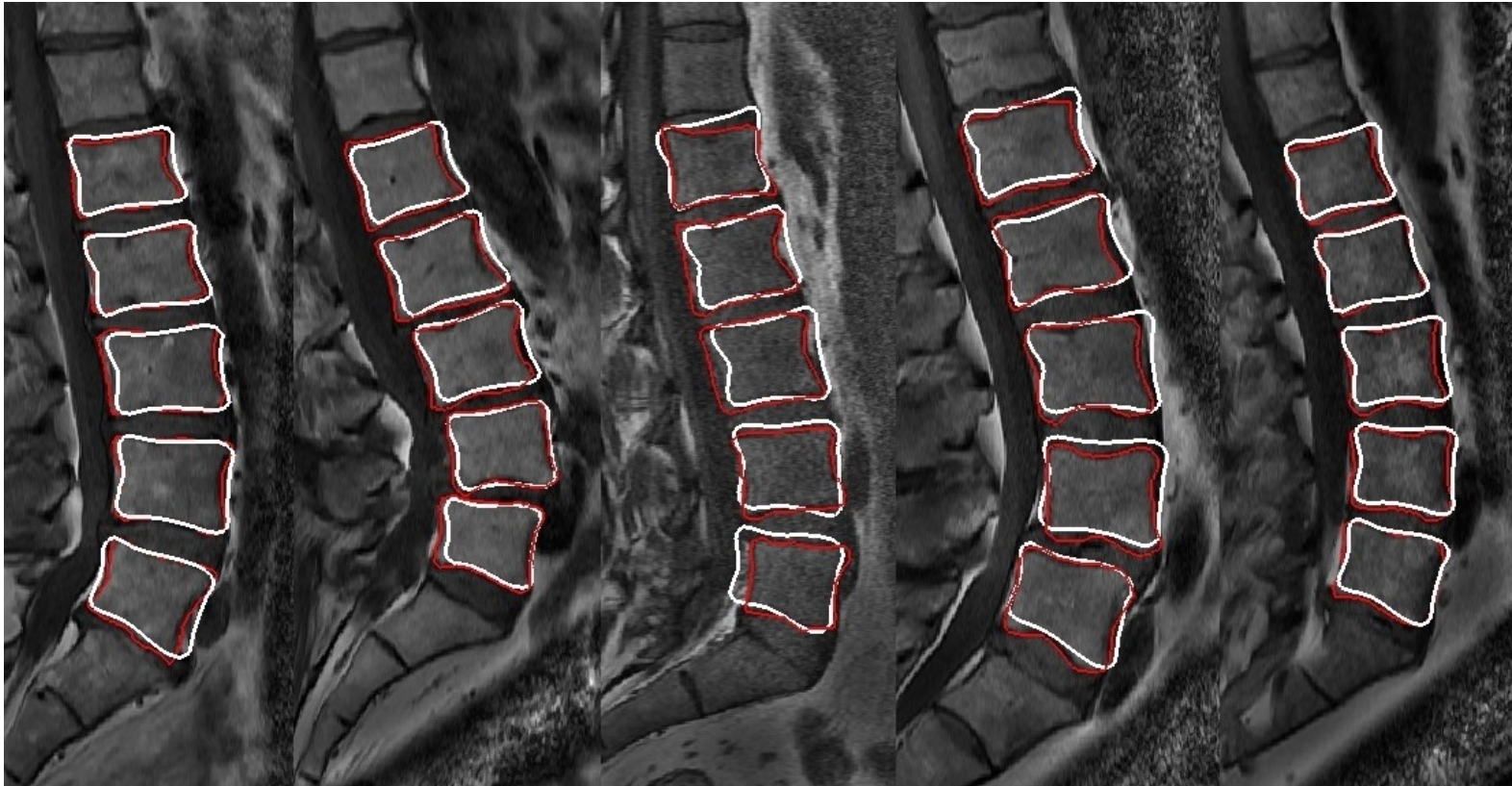
- Segmentation: 3D mean surface error $\cong 2.7 \pm 0.9$ mm.

Localization Results



Examples of localization and identification results.

Segmentation Results



Examples of segmentation result.

— Our segmentation
— Manual segmentation

Conclusion

- A method is proposed for Automatic localization, identification, and segmentation of vertebral bodies in volumetric MR images.
- Future work includes
 - Better evaluation on a large dataset of pathological cases.
 - Evaluation on other modalities like CT and Ultrasound.
 - Segmentation or sub-anatomical labeling of whole vertebrae.

Acknowledgement



Thank you ...

ece

Electrical and
Computer
Engineering



a place of mind