

Ínría



TIMC

3D inference of the spine from a depthmap of the back



¹Inria Morpheo, LJK, UGA, CNRS ²Anatoscope ³TIMC-SPM, CHU-Grenoble-Alpes, UGA









Research question

Can we guess scoliosis from a single depthmap?

Take away message

Depthmaps allow:

Scoliosis severity (°)

100

75

 Non-ionizing detection of scoliosis Location and quantification of the deformities Automatic characterization in 3D

Architecture of the regression model

 PCA^{-1} Inv. transformation of the PCA model MSEMean Squared Error $D_i \in \mathbb{R}^{224 \times 224}$ Depthmap of subject *i* $\theta_i \in \mathbb{R}^{20}$ Reduced spine representation $\Theta_i \in \mathbb{R}^{3 \times 17}$ 3D coordinates of the vertebrae $\omega_{ heta} \in \mathbb{R}$ Weight on the θ predictions $\omega_\Theta \in \mathbb{R}$ Weight on the Θ predictions

 $Loss = \omega_{\theta} MSE(\hat{\theta}, \theta) + \omega_{\Theta} MSE(PCA^{-1}(\hat{\theta}), \Theta)$

Adolescent Idiopathic Scoliosis

Dataset

121 subjects (31% with scoliosis) from different sources:

| Grenoble Hospital | NMDID [2] |
|--------------------------------------|--|
| 3D avatars using anatomical modeling | Segmentation of CT-Scans |
| | |

References

- [1] Choi et al. CNN-based Spine and Cobb Angle Estimator Using Moire Images. IIEE transactions on image electronics and visual computing, 5(2):135–144, 2017.
- [2] Edgard et al. New Mexico Decedent Image Database, Office of the Medical Investigator, University of New Mexico, 2020.
- [3] He et al. Deep residual learning for image recognition. CVPR 2016, 2016.
- [4] Kokabu et al. An algorithm for using deep learning convolutional neural networks with three dimensional depth sensor imaging in scoliosis detection. Spine Journal, 21:980-987, 2021.
- [5] Watanabe et al. An application of artificial intelligence to diagnostic imaging of spine disease: Estimating spinal alignment from moiré images. Neurospine, 16(4):697–702, 2019.
- [6] Yang et al. Development and validation of deep learning algorithms for scoliosis screening using back images. Communications Biology, 2:1–8, 2019.

| Metrics | | | | | | | | | | Classificatior | | | |
|---------|-------|----------------|--|-----------|-----------|----------------|------|------|-------------------|----------------|---------------------------|--|--|
| | | Positions (mm) | Angles (°) | | | Classification | | 60 - | ••••• | 10° threshold | | | |
| Method | Image | 3D | Sev | Кур | Lor | Sens | Spec | AUC | 50 - | | Regression (R=0.83 y=x | | |
| Ours | Depth | 7.1 (4.7) | 5.5 (6.2) | 6.3 (5.4) | 8.2 (6.9) | 64 | 99 | 90 | ~ 40 - | • | , NMDID cases | | |
| [6] | RGB | × | × | × | × | 88 | 84 | 95 |) uc | • | GH cases | | |
| [5] | Moiré | × | 3.4 (2.6) | × | × | NA | NA | NA | - 06 <u>i</u> ti | | | | |
| [/] | Donth | × | $\left[\Lambda \Lambda - \Lambda 7 \right]$ | × | × | 99 | 42 | ΝΙΔ | edi | | > | | |

Comparison with state-of-the-art reported values on different datasets. In positions: average distance error (with standard deviation). In curvatures: mean absolute error (with std) of severity, kyphosis and lordosis. In classification: sensitivity, specificity and AUC. NA: Not Available. X: Not computed.

nicolas.comte@inria.fr