





# Morphological risk of the femur causing anterior cruciate ligament (ACL) injuries

: Comparative 3-Dimensional Mean model analysis of ACL and non-injured patients

So-Heun Lee, M.S-PhD student <sup>5</sup>, Min Jung MD,PhD <sup>1,3</sup>, Kwang-Ho Chung MD,PhD <sup>1,4</sup>, Hyun-soo Moon MD <sup>1,2</sup>, PhD, Se-Han Jung, MD. <sup>1,3</sup>, Byeog-Hun jeon, MD <sup>1,3</sup>, Sung-hwan Kim,MD. PhD. <sup>1,3</sup>.

1. Arthroscopy and Joint Research Institute, Yonsei University College of Medicine, Seoul, Republic of Korea.

2. Department of Orthopedic Surgery, Gangnam Severance Hospital, Yonsei University College of Medicine, Seoul, Republic of Korea.

3. Department of Orthopedic Surgery, Severance Hospital, Yonsei University College of Medicine, Seoul, Republic of Korea.

4 Department of Orthopedic Surgery, Yongin Severance Hospital, Yonsei University College of Medicine, Yongin, Republic of Korea.

5. Medical Device Engineering and management, Yonsei University

## **Disclosures**

The authors do not have a financial interest or other relationship with a commercial company or institution.

✓ The authors do not have any affiliations or conflict of interest notifications to disclose.

## **Background & Purpose**

- ✓ Anterior cruciate ligament (ACL) injuries commonly affect young, active individuals and increase the risk of osteoarthritis (OA).
- ✓ However, there is inconsistency in the 2D radiographic findings of ACL injuries reported in the literature.
- ✓ This study directly compared statistical shape modeling (SSM) and principal component analysis (PCA) of femoral models between patients with and without ACL injury, and additionally developed a machine learning model to predict ACL injury based on femoral morphology.

## **Methods**

✓3D femoral shape simulations were retrospectively performed in 60 patients with ACL(male:30, female:30) and 60 matched controls(male:30, female:30).

√ The mean follow-up CT period: 3 days.

Variables	Normal	ACL	
Age, yrs	37.417±12.310	36.717±13.088	
Sex (male/female), n (%)	30 (50) / 30 (50)	30 (50) / 30 (50)	
Height	168.275±8.353	167.045±8.224	
Kg	69.203±15.947	67.992±13.565	
Affected side (Right/left), n (%)	37 (61.67) / 23 (38.33)	35 (58.33) / 25 (41.67)	

## **Methods**

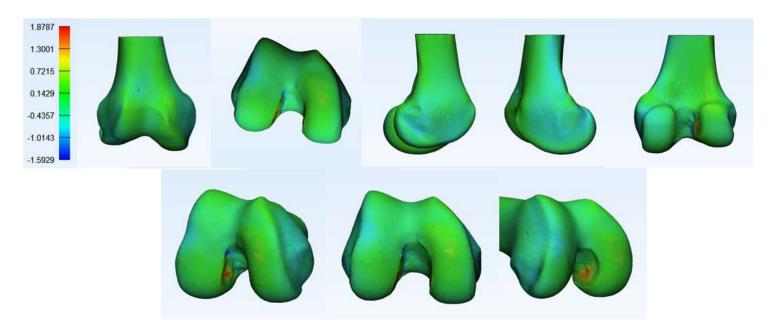
- Clinical assessments were performed using patient-reported outcomes(PROs).
- ✓ The 3D Visualize surface differences by distance map associated with patellar biomechanics were performed
  - ✓ Population 60 ACL vs 60 Normal
  - ✓ Male 30 ACL vs 30 Normal
  - √ Female 30 ACL vs 30 Normal
- ✓A logistic regression analysis was performed to identify key features that distinguish ACL-injured from non-injured subjects, which were subsequently used to train a *machine* learning model for ACL classification.



## Result Population ACL vs Non-ACL

### 3D Overview

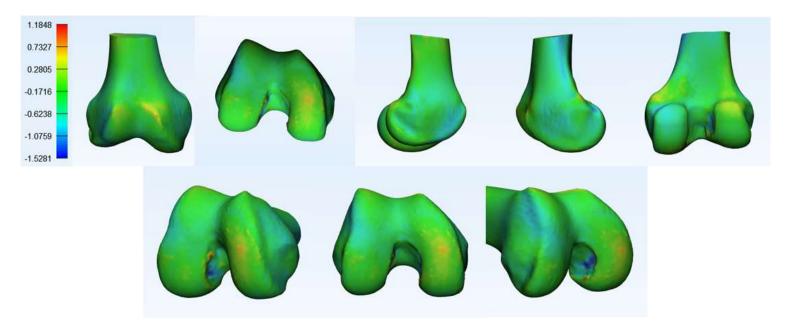
- Medial-lateral direction, both Intercondylar notch wall moved inward
- Anterior-posterior direction, trochlea anteriorized
- Proximal- distal direction, medial condyle distalized, lateral codyle proximalized



## Result Male ACL vs Non-ACL

### 3D Overview

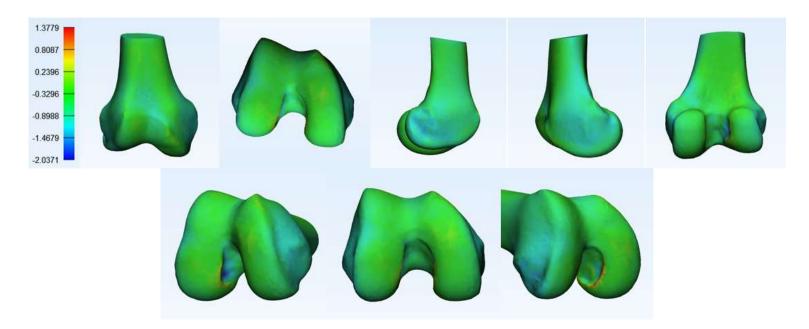
- Medial-lateral direction, both notch wall moved outward, femoral shaft lateralized
- Anterior-posterior direction, trochlea anteriorized
- Proximal- distal direction, Medial condyle distalized, lateral codyle proximalized



## Result Female ACL vs Non-ACL

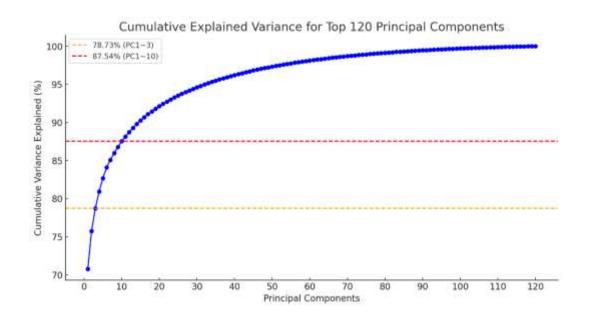
### ■ 3D Overview

- Medial-lateral direction, both notch ridges moves inward, and epicondyle shrinked
- Anterior-posterior direction, patella surface shows decreased concavity
- Proximal- distal direction, medial condyle distalized, lateral condyle proximalized





## Results – logistic regression analysis



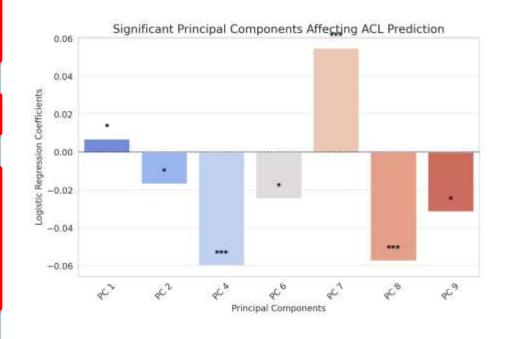
### ✓ Cumulative explained variation

- PC1~3 explained 78.73%, and PC1~10 accounted for 87.54% of population variance



## Results – logistic regression analysis

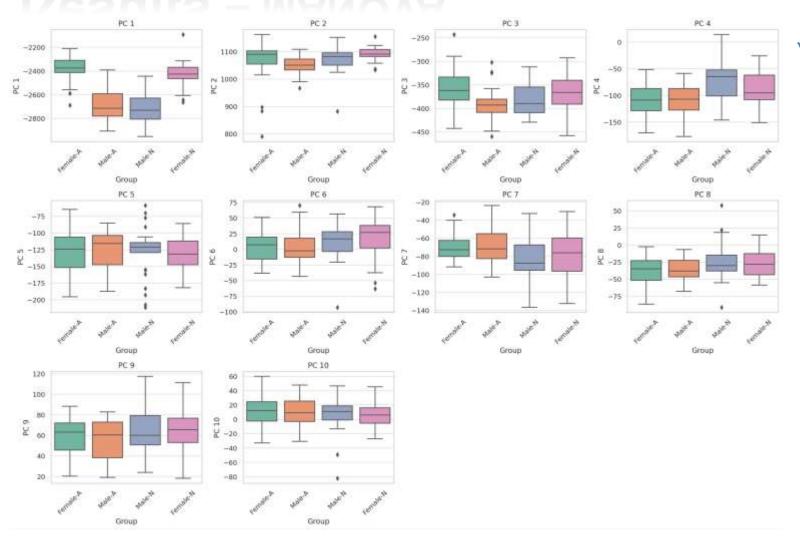
PC	RC	P-value	Adjusted RC	Adjusted P- value
PC1	-1.007	0.080	1.335	0.016
PC2	0.539	0.065	-0.861	0.024
PC3	1.335	0.016	0.403	0.305
PC4	-0.861	0.024	-2.049	0.000
PC5	0.403	0.305	0.130	0.641
PC6	-2.049	0.000	-0.672	0.036
PC7	0.130	0.641	1.267	0.001
PC8	-0.672	0.036	-1.240	0.000
PC9	1.267	0.001	-0.649	0.042
PC10	-1.240	0.000	0.605	0.072



- $\sqrt{\text{Significance markers:***}} (p < 0.001), ** (p < 0.01), * (p < 0.05)$ 
  - PC7 contributes significantly to ACL risk, and PC4 PC8 strongly reduce ACL risk.
  - Other PCs (PC1, PC2, PC6, and PC9) have moderate influences.



## Results - MANOVA

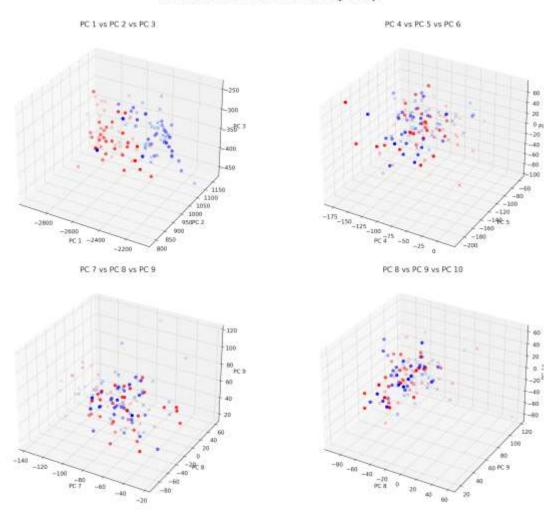


## Group Differences in ACL and Sex by MANOVA

- PC1 captures morphological features associated with ACL injury in males.
- PC2 also demonstrated noticeable differences between sexes,
- PC3 through PC10, no substantial differences were observed among the four groups

## Results – 3D PC scatter

3D PCA Scatter Plots: PC1-PC10 by Group



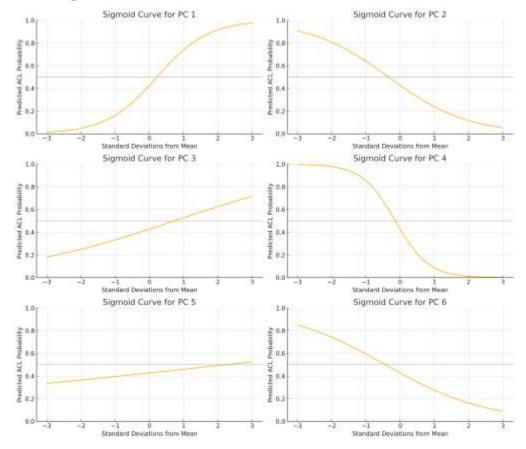
## ✓ Group Differences in ACL and Sex by MANOVA

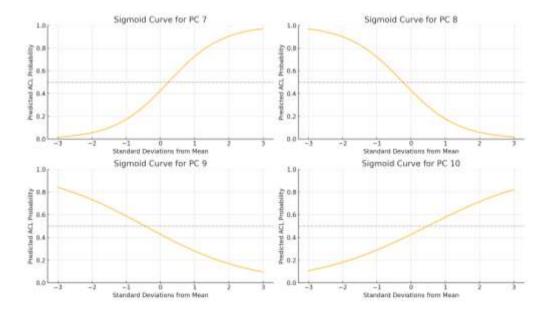
- PC1 Clearly separates ACLinjured and non-injured groups, especially in males
- PC2 Distinguishes between sexes regardless of ACL status.
- PC3 Shows minor separation, contributing only when combined with PC1.



## **Predictable ACL probabilty**

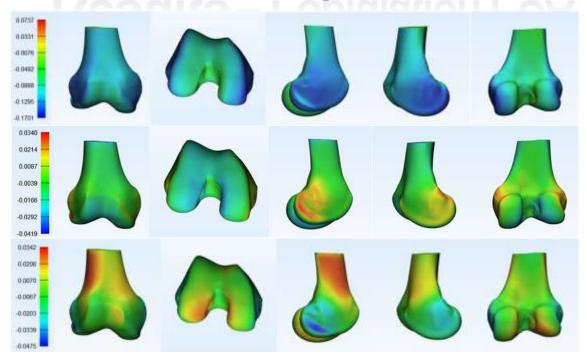
### Sigmoid curve for PCs

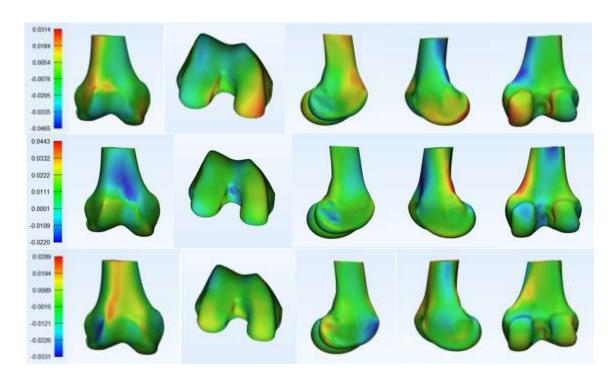






## **Results – Poplulation PCA**

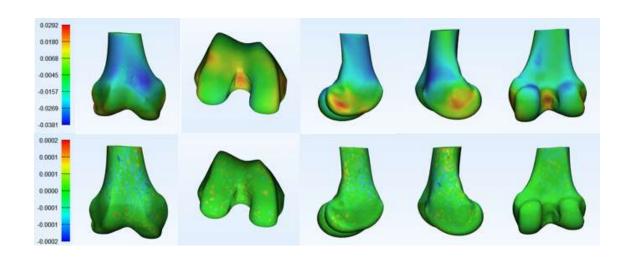


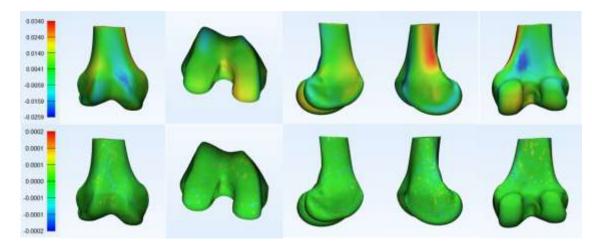


- ✓ PC1-3, PC6, PC7,PC9 are related to ACL
  - PC1 Associated with ACL-related structural variation, showing intercondylar notch narrowing, especially in ACL-injured males.
  - PC2 Reflects sex-based differences, , characterized by epicondylar shrinking
  - PC3 Captures rotational geometry and secondary geometric ACL risk factors to PC1.



## Results – Poplulation PCA





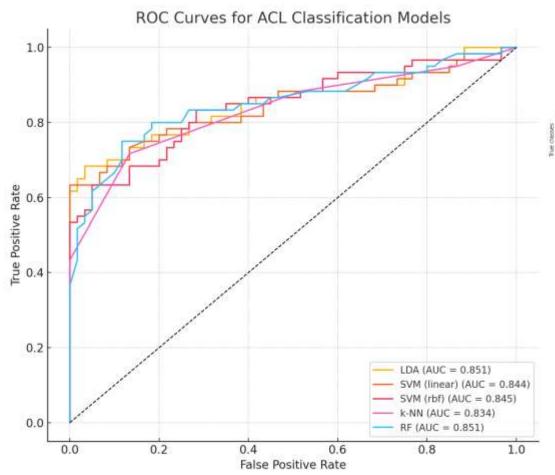
#### ✓ PC4, PC8 are related to non-ACL

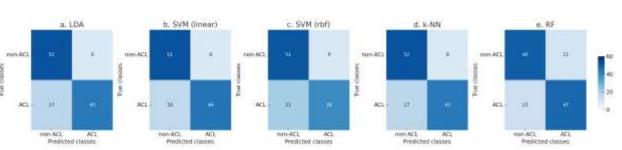
- close to the normal shape, their directions were statistically reversed to align with ACL-associated shape patterns
- There was no appreciable difference in numerical values between the adjusted and mean models.



### ML-based prediction model for ACL

### ■ Receiver Operating Characteristic(ROC) Curve





Classifier	Accuracy	Sensitivity	Specificity	AUC
LDA	0.792	0.717	0.867	0.851
SVM( with linear)	0.8	0.733	0.867	0.844
SVM( with rbf)	0.75	0.65	0.85	0.845
K-NN	0.792	0.717	0.867	0.834
Random Forest	0.8	0.783	0.817	0.851



## **Limitations & Future Directions**

- **✓ Limitations:** 
  - √ Small sample size
- ✓ Future Research:
  - ✓ Long-bone
  - √ tibia

## Conclusion

- Regardless of sex, femoral morphology characterized by condylar asymmetry, shaft anteriorization/lateralization, and notch narrowing may lead to rotational misalignment, increasing the risk of ACL injury
- ✓ <u>In males</u>, ACL injury is associated with <u>intercondylar notch narrowing</u>, *epicondylar shrinking*, which may reflect a tighter bony geometry and internally rotated femoral alignment, increasing anterior ligament strain.
- ✓ In females, ACL injury risk is linked to patellar anteriorization and adductor tubercle elevation, features that may correspond to a more externally rotated femoral alignment and altered patellofemoral tracking, contributing to ligament overload.



### References

- Pedoia, V., Lansdown, D., Zaid, M., McCulloch, C., Souza, R., , C., & Li, X. (2015). Three-dimensional MRI-based statistical shape model and application to a cohort of knees with acute ACL injury. Osteoarthritis and cartilage, 23 10, 1695-703. https://doi.org/10.1016/j.joca.2015.05.027.
- Polamalu, S., Musahl, V., & Debski, R. (2020). Tibiofemoral bony morphology features associated with ACL injury and sex utilizing three-dimensional statistical shape modeling. Journal of Orthopaedic Research®, 40, 87 94. https://doi.org/10.1002/jor.24952.
- Ghidotti, A., Regazzoni, D., Cohen, M., & Condello, V. (2024). A Statistical Shape Model of Femur of ACL-Injured Knee. Volume 2A: 44th Computers and Information in Engineering Conference (CIE). https://doi.org/10.1115/detc2024-142809.
- 4 Renato Andrade et al. Notch morphology is a risk factor for ACL injury: a systemic review and meta-analysis, Journal of ISAKOS, Volume 1, Issue 2, 2016, Pages 70-81, ISSN 2059-7754, https://doi.org/10.1136/jisakos-2015-000030.
- 5 Ambellan, F., Lamecker, H., Von Tycowicz, C., & Zachow, S. (2019). Statistical Shape Models: Understanding and Mastering Variation in Anatomy. Advances in experimental medicine and biology, 1156, 67-84. <a href="https://doi.org/10.1007/978-3-030-19385-0">https://doi.org/10.1007/978-3-030-19385-0</a> 5.
- Rolfe, S., Pieper, S., Porto, A., Diamond, K., Winchester, J., Shan, S., Kirveslahti, H., Boyer, D., Summers, A., & Maga, A. (2021). SlicerMorph: An open and extensible platform to retrieve, visualize and analyse 3D morphology. Methods in Ecology and Evolution, 12, 1816 1825. <a href="https://doi.org/10.1111/2041-210X.13669">https://doi.org/10.1111/2041-210X.13669</a>.
- Porto, A., Rolfe, S., & Maga, A. (2021). ALPACA: A fast and accurate computer vision approach for automated landmarking of three-dimensional biological structures. Methods in Ecology and Evolution, 12, 2129 2144. https://doi.org/10.1111/2041-210X.13689.

