

Abstract

This project investigates changes in mechanical properties of kidney cells when exposed to TGF- β 1, which is known to induce renal disease [1]. The aim of this project is to provide insight on the progression of diabetic nephropathy from a mechanical perspective based on changes in mechanical properties observed in single cells using atomic force microscopy.

Equations

$$F(\delta) = \frac{4}{3} \cdot \frac{E}{1 - \nu^2} \cdot \sqrt{R} \cdot \delta^{3/2} \quad (1)$$

The Hertz/Sneddon spherical indentation model (Eq. ??) is matched to the force indentation curve to find the apparent elasticity of an experiment.

F ::Force
 E ::Young's Modulus
 ν ::Poisson's Ratio
 R ::Indenter Radius
 δ ::Indentation depth

$$\hat{P}(G_2 | x) = \frac{P(x | G_2) \cdot P(G_2)}{P(x | G_1) \cdot P(G_1) + P(x | G_2) \cdot P(G_2)} \quad (2)$$

The Bayesian classifier is a function based on Bayes theorem that finds a posterior probability (the probability of a precondition given the result).

$P(G | x)$::Posterior
 $P(x | G)$::Likelihood
 $P(G)$::Prior
 $P(x)$::Evidence

Where the likelihood of a given group is determined by fitting the observed occurrences to a distribution / Probability Density Function (PDF). 3 distribution models are tested: Gaussian (Eq. ??), Skewed Normal (Eq. ??), and Kernel Density Estimation (Eq. ??).

$$\hat{P}(x | G) = \frac{1}{\sigma_G \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x - \mu_G}{\sigma_G} \right)^2} \quad (3)$$

$$\hat{P}(x | G) = \phi(x; \mu_G, \sigma_G) \cdot \Phi \left(\alpha_G \cdot \frac{x - \mu_G}{\sigma_G} \right) \quad (4)$$

$$\hat{P}(x | G) = \frac{1}{nh} \sum_{i=1}^n K \left(\frac{x - x_{iG}}{h} \right) \quad (5)$$

$\hat{P}(x | G)$:: Group Probability Density Function

x :: Observation (i.e. Young's Modulus)

σ_G :: Group Standard Deviation

μ_G :: Group Mean

$\phi(x; \mu, \sigma)$:: Normal PDF evaluated at x

α_G :: Group Skew Parameter

$\Phi(z)$:: Standard Normal CDF

x_{iG} :: Observed Data Points from Group G

n :: Number of Observations

$K(\cdot)$:: Kernel Function (i.e. Gaussian)

h :: Bandwidth (Smoothing Parameter)

Recommendations

► Repeat with larger experimental dataset
Applying the methodology described on larger datasets would allow for a more robust estimate with more extensive testing and validation.

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Introduction

This project investigates the predictive power of renal tubular epithelial cell stiffness as a biomarker for the progression of Diabetic Nephropathy (DN). DN is a common and serious complication of diabetes resulting in kidney failure due to progressive damage to the nephrons, the functional units of the kidney responsible for filtering the blood [2]. This loss of function is due to physical changes at the cellular level induced by cytokine TGF- β 1 associated with an observable change in cytoskeleton stiffness [3].

A force against indentation curve of a cells can be observed using Atomic Force Microscopy (AFM) where the deflection of a very fine probe on a flexible cantilever is measured to detect contact forces. From the spring constant of the cantilever the indentation and force exerted can be determined as the assembly is advanced into the sample. This curve can then be fitted against an elastic deformation model to determine an apparent Young's Modulus (YM).

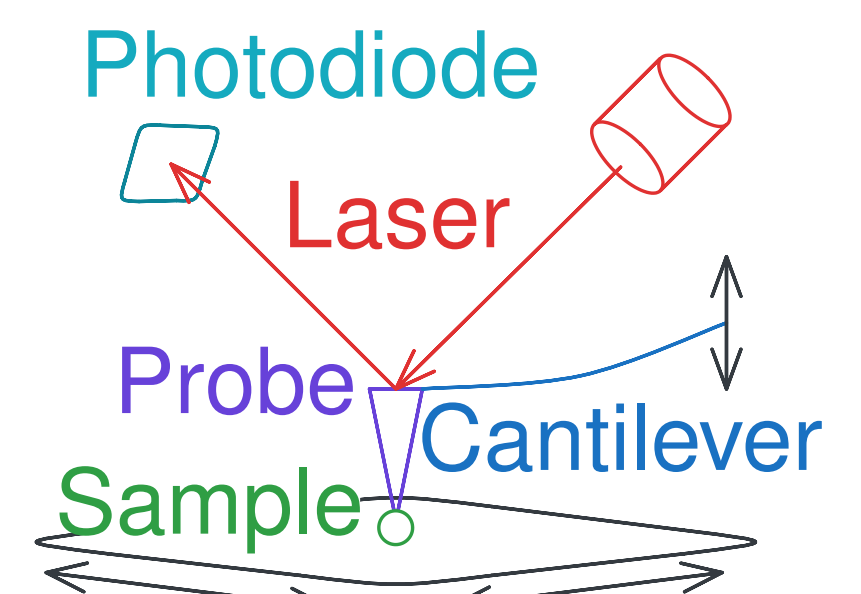


Fig. 1: AFM Diagram

The probability a given cell is healthy or diseased can be predicted from the observed distributions of YM of cells that have not been exposed to TGF- β (Control) and those that have (Treated) by a Bayesian classifier.

Methodology

Observe Cell Response

- Single cell indentation tests via atomic force microscopy X5 per Cell
- Pre-processing raw data to force vs indentation depth curves

Elasticity Modeling

- Estimate YM via for each test by fitting observed response to an indentation model
- Estimate apparent YM for each cell and account for uncertainty and error

Determine Effect Strength

- Estimate healthy vs diseased group characteristics, and uncertainty
- Quantify statistical significance and predictive power of the observed effect

Construct Classifier

- Determine suitable likelihood probability density functions
- Construct Bayesian classifiers and assess performance

Results

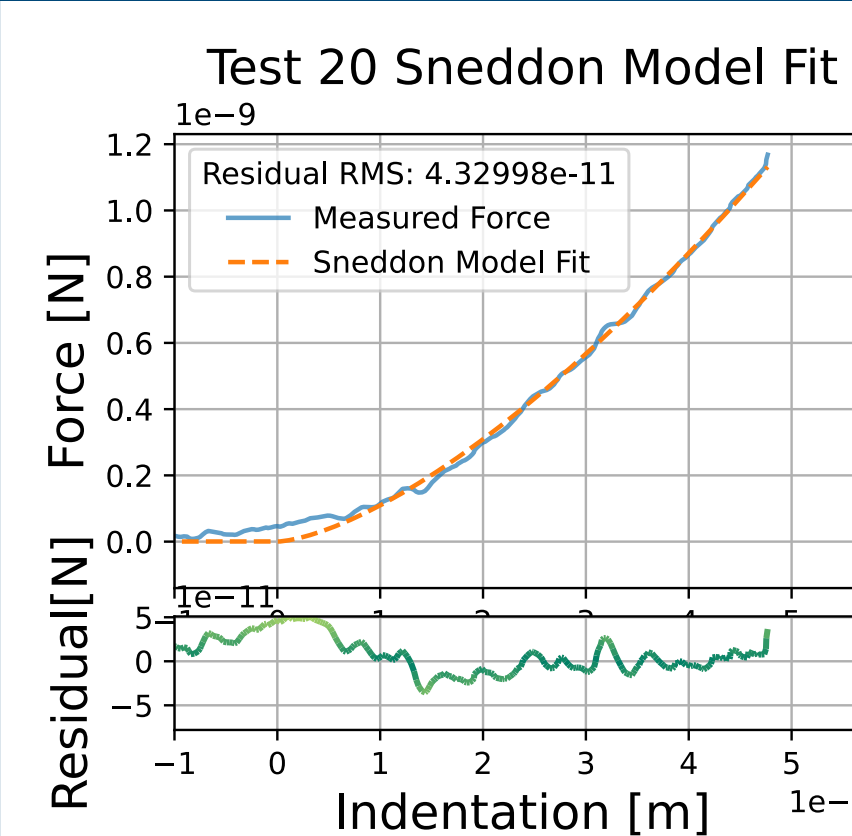


Fig. 2: Indentation Fit

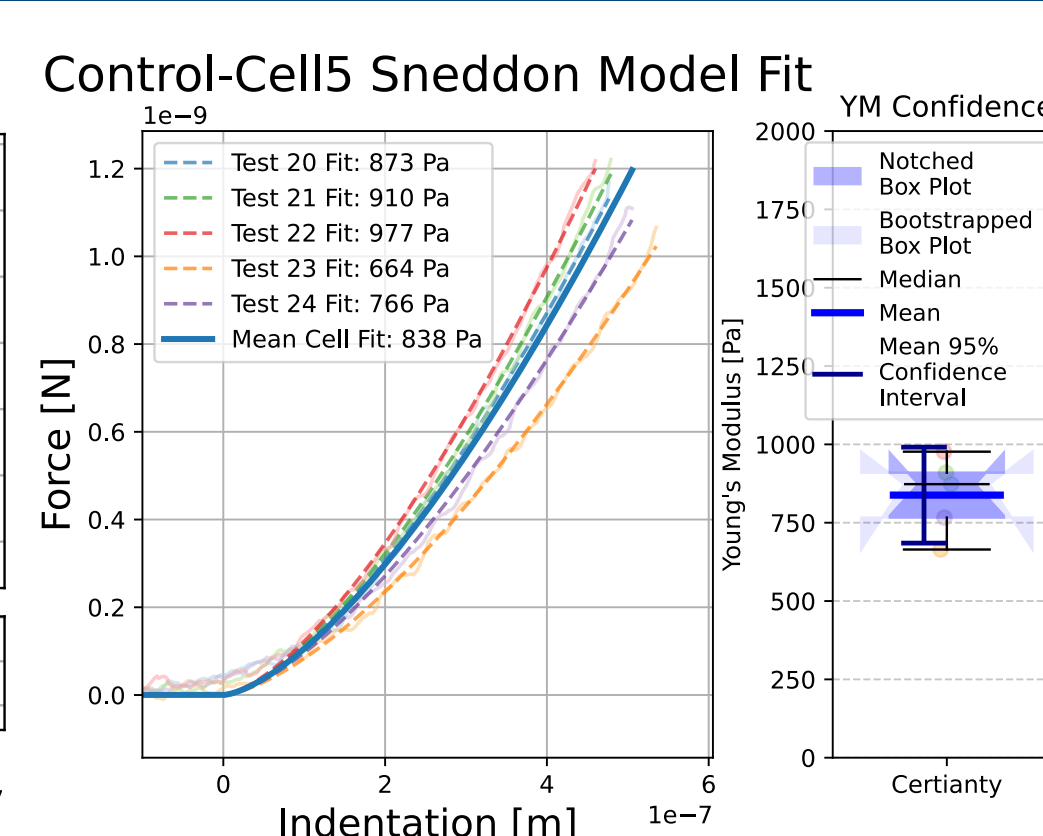


Fig. 3: Cell Elasticity

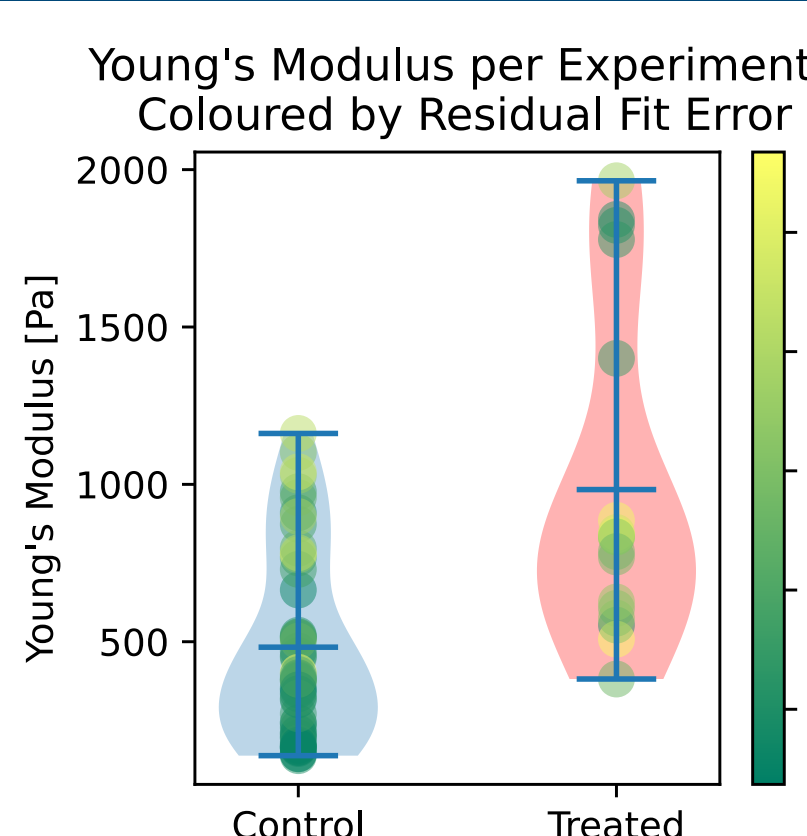


Fig. 4: Observed Effect

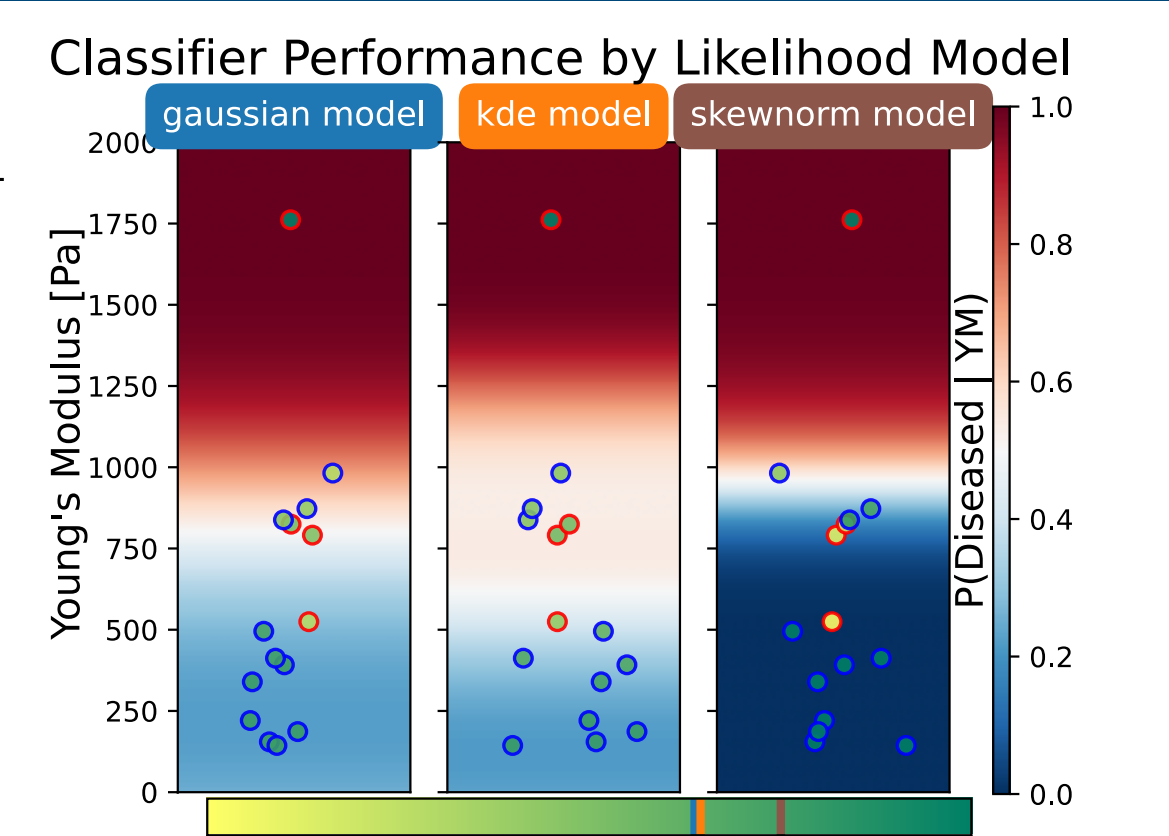


Fig. 5: Classifiers

The models achieve high average accuracy with just a few samples but are significantly more reliable with more cells. The skewed normal is the most effective distribution model averaging >99% accuracy over 50,000 trials with 5 sample cells.

Discussion

Cell relaxation observed over successive tests

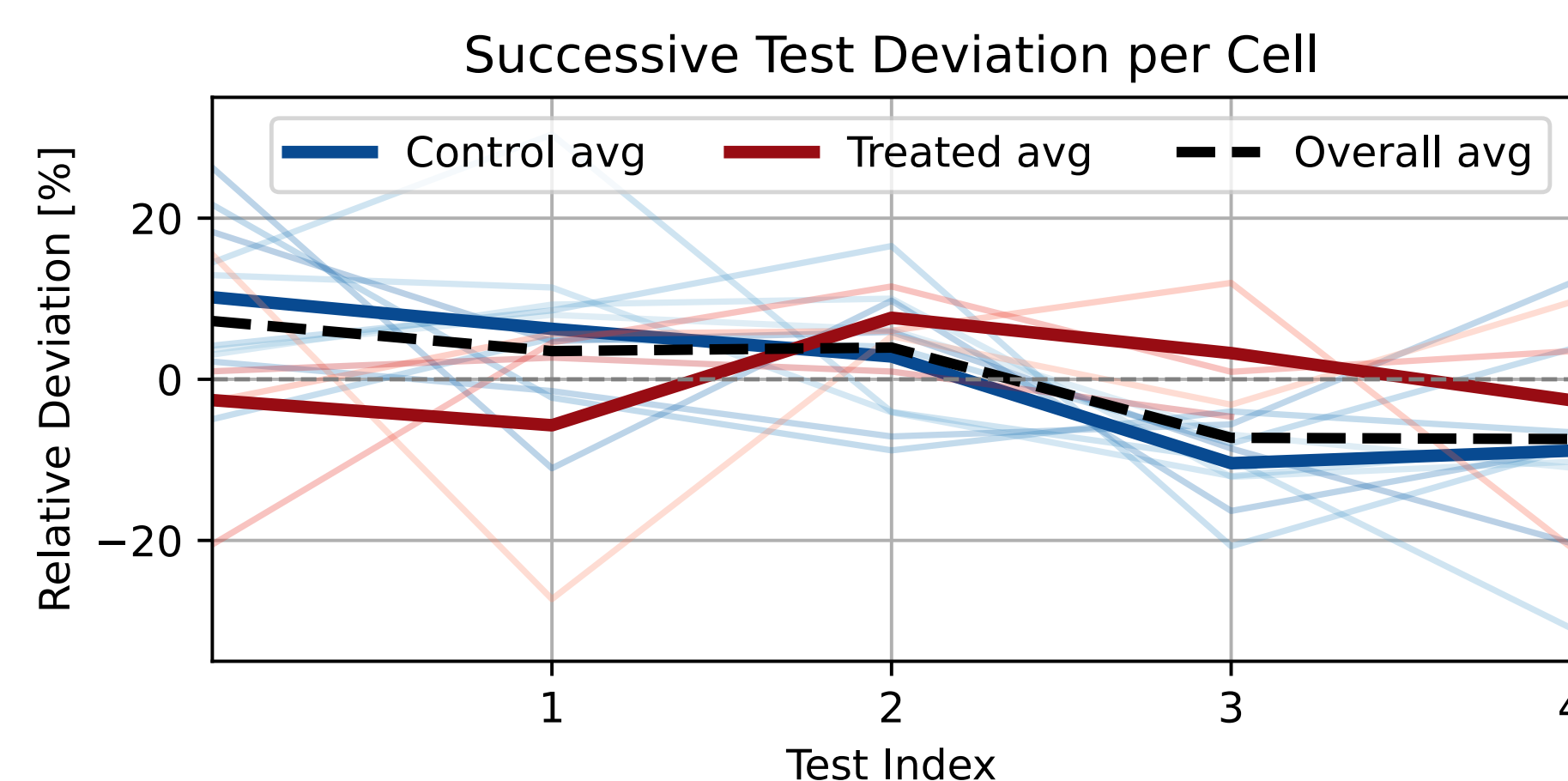


Fig. 7: Cell Relaxation Trend

Confidence not sufficient to guarantee effect

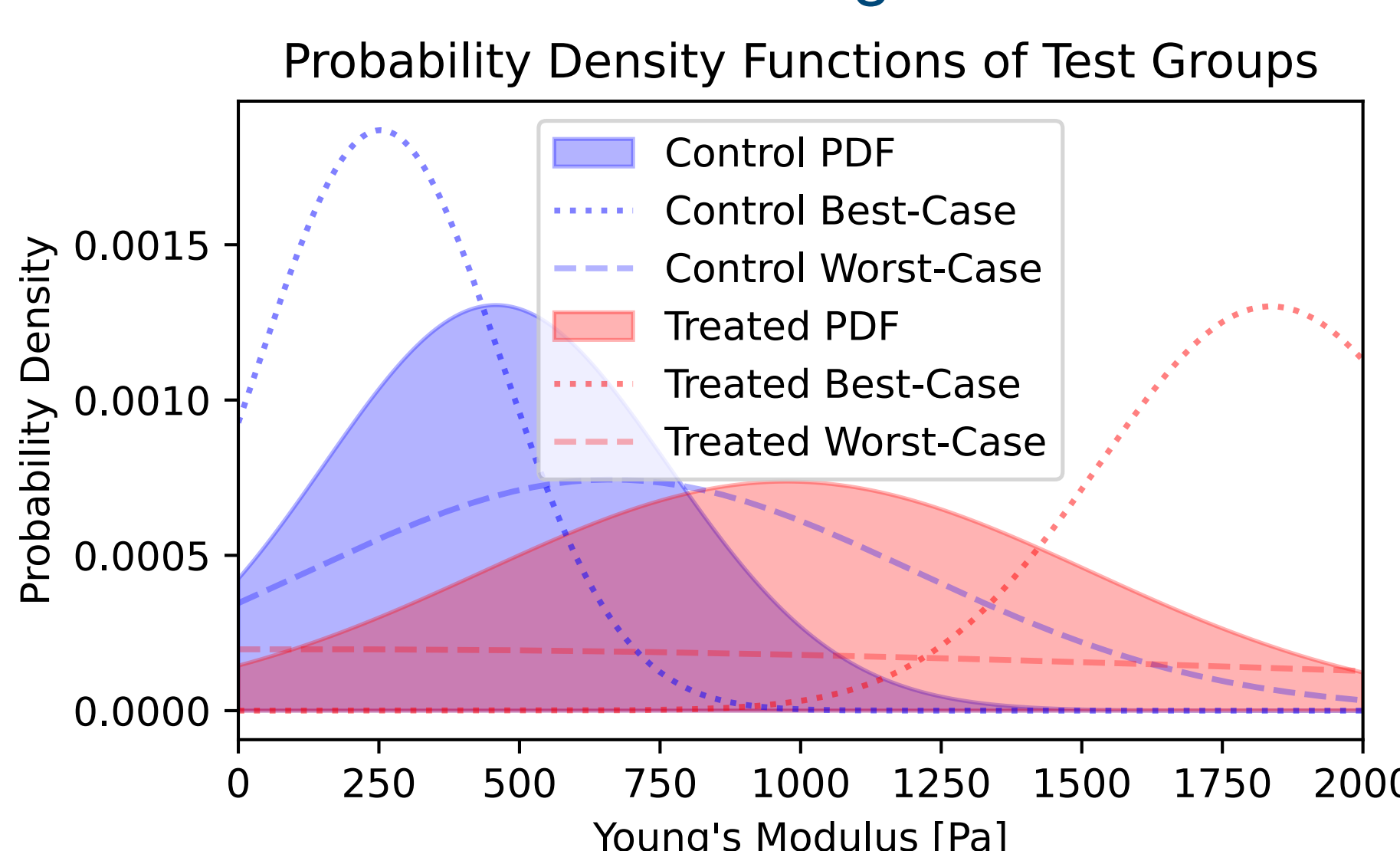


Fig. 8: 95% Confidence PDF Limit Cases

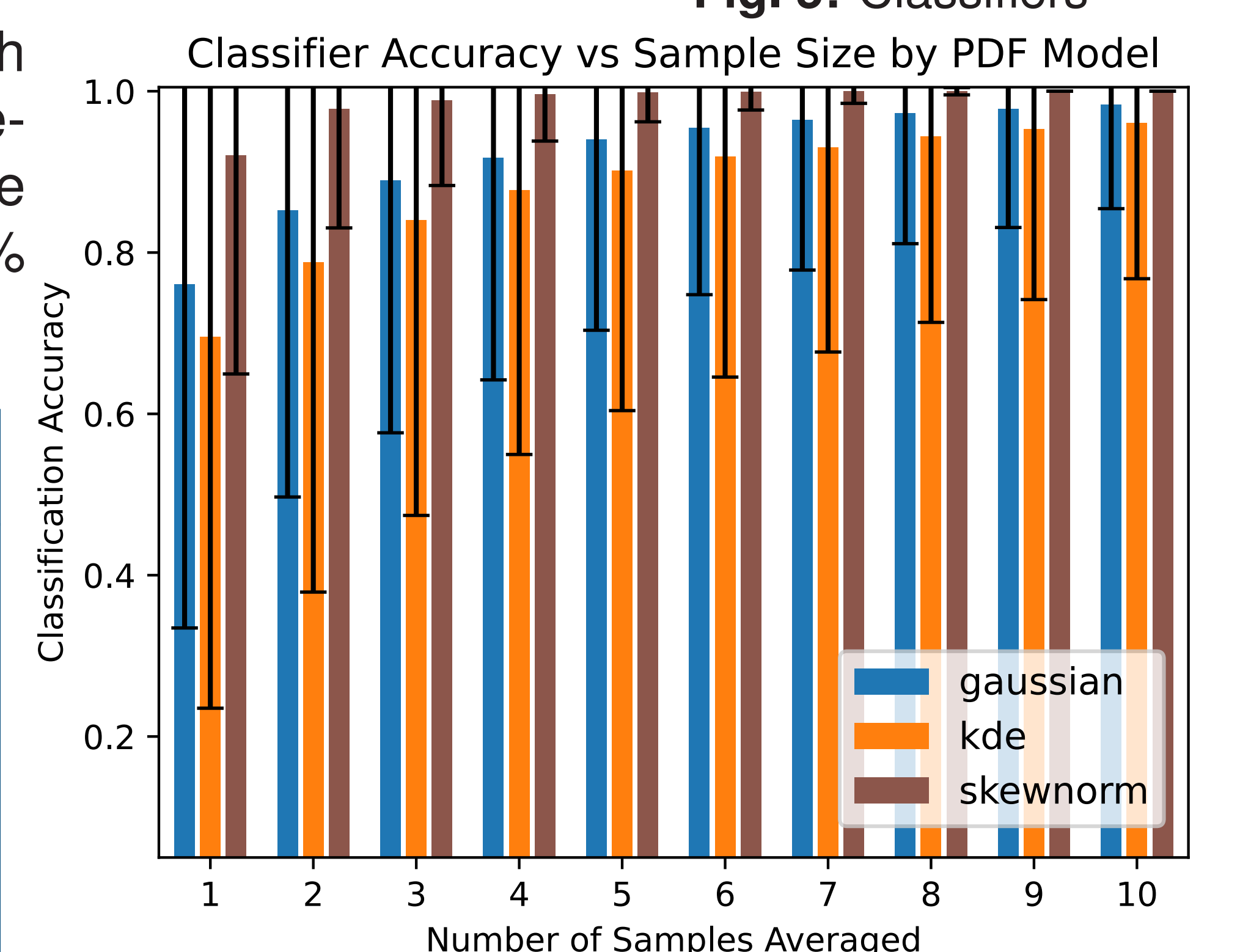


Fig. 6: Performance for 50,000 simulated trials

References

- [1] M. E. Gentle, S. Shi, I. Daehn, *et al.*, "Epithelial Cell TGF Signaling Induces Acute Tubular Injury and Interstitial Inflammation," *Journal of the American Society of Nephrology* : JASN, vol. 24, no. 5, pp. 787–799, Apr. 30, 2013, ISSN: 1046-6673. DOI: 10.1681/ASN.2012101024. PMID: 23539761. (visited on 02/04/2025).
- [2] W. Metcalfe, "How does early chronic kidney disease progress?: A Background Paper prepared for the UK Consensus Conference on Early Chronic Kidney Disease," *Nephrology Dialysis Transplantation*, vol. 22, pp. ix26–ix30, suppl_9 Sep. 1, 2007, ISSN: 0931-0509. DOI: 10.1093/ndt/gfm446. (visited on 01/29/2025).
- [3] C. E. Hills, E. Siamantouras, S. W. Smith, P. Cockwell, K.-K. Liu, and P. E. Squires, "TGF modulates cell-to-cell communication in early epithelial-to-mesenchymal transition," *Diabetologia*, vol. 55, no. 3, pp. 812–824, Mar. 1, 2012, ISSN: 1432-0428. DOI: 10.1007/s00125-011-2409-9. (visited on 01/29/2025).