Learning-Based Animation of Clothing for Virtual Try-On

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~80% of clothing is purchased at brick-and-mortar stores



Images from Wikimedia Commons



Virtual try-on



Garment





Different bodies



Garment



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Virtual try-on





Virtual try-on Key insights

- Cloth deformations are highly nonlinear
 - Physics-based models
 - Data-driven models (e.g., neural networks)
- Clothing deformations are strongly correlated with body shape and motion
- The human body can take a wide ranges of shapes



Human model



 $\boldsymbol{\theta} \in \mathbb{R}^{69}$ (Pose coefficients) $\boldsymbol{\beta} \in \mathbb{R}^{10}$ (Shape coefficients)

SMPL [Loper et al. 2015]



Human model





Physics-based models



Cirio et al. 2015



Narain et al. 2012



Tang et al. 2018



Physics-based models



Tang et al. 2018

Narain et al. 2012











Kavan et al. 2011



Hahn et al. 2014

Wang et al. 2010



Guan et al. 2012





Hahn et al. 2014

Guan et al. 2012





DRAPE Guan et al. 2012

- Preserves the style across different body shapes
- The size of the garment changes



ClothCap Pons-Moll et al. 2017

- Uses retargeting
- The wrinkles are copied from one shape to another





DRAPE Guan et al. 2012



ClothCap Pons-Moll et al. 2017





Our method

Our method



250 fps



Our method Key ideas

- 1. Build a rich dataset of dressed character simulations
- 2. Learn deformations in pose-space
- 3. Disentangle two sources of deformation:
 - Static fit deformations (shape dependent)
 - Dynamic wrinkle deformations (shape and motion dependent)
- 4. Capture temporal dependencies





Data-generation pipeline





Data-generation pipeline



Dataset 56 sequences × 17 shapes



ARCSim [Narain et al. 2012]



Data preprocessing



Simulated mesh



Data preprocessing





Data preprocessing



















Step 2: Δ_{wrinkle}







- Modelling temporal dependencies:
 - Add the output of the previous frame as input (e.g., De Aguiar et al. SIGGRAPH 2011, Casas et al. i3D 2018)
 - Recurrent neural networks (e.g., RNN, GRU, LSTM)



Postprocessing



Before

After



Qualitative evaluation

Shape generalization



Cloth simulation

Retargeting

Our method



Qualitative comparison



ClothCap (Retargeting) [Pons-Moll et al. 2017]

Our method



Qualitative comparison



ClothCap (Retargeting) [Pons-Moll et al. 2017]

Our method



Qualitative comparison





Qualitative evaluation

Pose generalization





Contributions

Clothing deformations as a function of body shape and motion

- Captures highly nonlinear effects
- Very fast clothing animations: 250 fps!

Future work

Generalize to multiple garments

Future work

Generalize to multiple garments

Collisions

Future work

Generalize to multiple garments

Collisions

Loose garments

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