

# Sackcloth or Silk? The Impact of Appearance vs Dynamics on the Perception of Animated Cloth

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**Figure 1:** Different fabrics have both different visual appearance and mechanical properties. We create replicas of several common woven fabrics, like the cotton or silk shown in the image, covering a wide range of movements in a set of video stimuli. Then, we combine the appearance of each fabric with the dynamics of the other ones and vice versa, and perform psychophysical experiments to study the relative importance of appearance and dynamics when perceiving cloth.

## Abstract

Physical simulation and rendering of cloth is widely used in 3D graphics applications to create realistic and compelling scenes. However, cloth animation can be slow to compute and difficult to specify. In this paper, we present a set of experiments in which we explore some factors that contribute to the perception of cloth, to determine how efficiency could be improved without sacrificing realism. Using real video footage of several fabrics covering a wide range of visual appearances and dynamic behaviors, and their simulated counterparts, we explore the interplay of visual appearance and dynamics in cloth animation.

## CR Categories:

**Keywords:** perception, cloth, appearance, dynamics, distance

## 1 Introduction

3D animation is becoming more and more sophisticated. With the evolution of rendering algorithms, motion capture techniques and physics simulators, new productions progressively offer more complex shots and more stunning visuals. However, it is often the case that intricately modeled details and complex simulations are employed to create scene elements that may go unnoticed by the viewer, which is not a very efficient use of resources.

This leads to the following question, which we aim to investigate

in this paper: *Do all elements of a simulation need to be physically correct in order to achieve realism?* Given the very large space of possible parameters, we focus here on a very common scenario where physically-based simulations are employed in current 3D application areas: the rendering and animation of photo-realistic cloth. In particular, we analyze the interplay of visual appearance and dynamics and how it affects the viewer. The goal is to analyze when (and if) a simplified simulation can be used in the presence of a very accurate shader, or vice versa. Do both appearance and dynamics need to be perfectly simulated in order to convey the desired impression? Can different strategies be employed depending on the particular types of fabric being depicted?

To answer these questions, we first captured videos of seven different real cloth samples made of different fabrics covering a wide range of visual appearances and dynamic behaviors. We also created photo-realistic synthetic versions that emulated the real cloth samples as closely as possible. Given these seven ground-truth animations, we rendered all possible combinations of appearance and dynamics, yielding a 7x7 stimulus matrix where only the diagonal elements had matching characteristics. We then conducted two perceptual experiments, where participants were asked to match these stimuli with the ground-truth filmed videos, and were also asked to identify which animation had mismatching motion and appearance properties.

To our knowledge, this is the first effort towards understanding the relative weightings of appearance and dynamics on the perception of photo-realistic animated cloth. Although we focus here on the particular case of cloth simulation, our methodology could be extended to other scenarios. Our results may be useful to guide a better distribution of resources when planning shots involving cloth simulations, or could affect how shot approvals are done. For instance, if the perception of a given fabric is strongly influenced by its visual appearance and less by its dynamics, then viewing the simulation without a reasonable depiction of the final shader to be employed, and vice versa, would not be sufficient to predict the final result.



**Figure 2:** Comparison between the real fabrics and the CG replicas. From left to right: burlap, canvas, denim, linen, cotton, polyester satin and sheer silk. The images show renders, the insets are close up pictures of the real fabrics in the case of the first five rows. In the case of the last two fabrics on the right (polyester satin and sheer silk), the weaving pattern is too small to notice at normal viewing distances. Thus, for polyester satin the inset shows the fabric wrapping a cylinder along the warp and weft directions to show the viewing and lighting dependent anisotropic highlights. For the sheer silk, the inset shows the real fabric draping the swivel stool.

## 2 Related Work

Perceptually-based computer graphics is an active research field. The key idea is to take into account the limits of the human visual system to improve the efficiency of realistic image synthesis and animation. We refer the interested reader to the many existing surveys and courses (e.g., [O’Sullivan et al. 2004; Bartz et al. 2008; McNamara et al. 2011]), and focus here on appearance and dynamics.

**Appearance** Many approaches focus on generating visually plausible materials. Pellacini et al. [2000], Westlund and Meyer [2001] and Ferwerda et al. [2001] developed psychophysically-based models for gloss perception. Wills et al. [2009] performed similar experiments to derive a perceptual space of measured BRDFs. Vangorp et al. [2007] evaluated the influence of shape and illumination on surface gloss perception, showing how objects with smooth bumps provide more cues than simpler ones like spheres. Other studies include translucency and subsurface scattering [Fleming and Bülthoff 2005; Gkioulekas et al. 2013], or surface texture and reflectance [Dana et al. 1999; Filip et al. 2008; Jarabo et al. 2014]. Fleming and colleagues [2001; 2003] conducted reflectance matching experiments to demonstrate that people can recognize material properties more accurately under natural illumination than under artificial lights. Other examples focus on perceptually guided global illumination [Myszkowski 2002; Stokes et al. 2004]. Ramanarayanan and colleagues [2007; 2008] evaluated the effects of changes in environment lighting over different shapes and materials. Through several transformations in the illumination maps, such as warping or blurring, they found that many objects had the same appearance (they are visually equivalent) when illuminated by both transformed and original maps. Similar studies evaluated the effect of approximations in illumination on the perception of complex animated scenes [Jarabo et al. 2012] or materials [Křivánek et al. 2010].

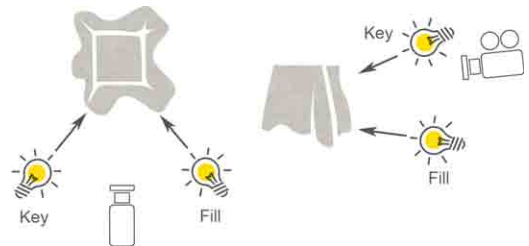
**Dynamics** Some studies have evaluated the effects of degrading or distorting physically-based simulations on the perceived plausibility of animations, e.g., [O’Sullivan et al. 2003; Yeh et al. 2009; Han et al. 2013]. Similar studies have also been conducted in the context of cartoons [Garcia et al. 2008]. Other works focus on collisions; O’Sullivan et al. [1999] developed a model of collision perception for real-time animation, while Dingliana and O’Sullivan [2000; 2001] examined the perception of detail simplifications for LOD rigid-body physically-based animation. Some other works evaluate the perception of dynamics on animated characters. Reitsma et al. [2003] studied the visual tolerance of ballistic motion for character animation, finding that horizontal velocity errors are more detectable than vertical. Vicovaro et al. [2012] evaluated the plausibility of altered throwing motions.

Finally, Hoyet et al. [2012] conducted several psychophysical experiments to measure the perceived realism of pushing interactions, evaluating the influence of timing errors or force mismatches.

Two previous studies are relevant to our work. McDonnell et al. [2006] evaluated the perceptual impact of different geometric and image-based LOD representations of animated cloth, and guidelines for developing crowd systems with realistic clothed humans were presented. Most recently, Sigal et al. [2015] developed a perceptual control space for cloth dynamics, mapping the complex parameters from any physical simulator to a few intuitive and meaningful parameters learned from a set of perceptual experiments.

## 3 Stimuli Creation

In order to cover a reasonable range of different fabric appearances and dynamics, we chose seven commonly used woven cloths. In approximate order of more to less stiff, the selected fabrics are: Burlap (also commonly known as Sackcloth), Canvas, Denim, Linen, Cotton, Polyester satin and sheer Silk. We acquired real samples of all of them, cut into squares of 1x1 meters. They all are of roughly the same albedo, in order to avoid color being a confounding factor for the experiments (see Figure 2).



**Figure 3:** Lighting studio setup for capturing the video footage of the real cloth samples, from bottom and side views.

We then recorded videos of all the fabrics in a studio with diffuse black walls, floor, and roof, using two spot lights placed at about 45 degrees from the focal plane (Figure 3). Every piece of cloth was recorded while draping over a flat swivel stool which then spins, in order to show as many mechanical and dynamic properties of the fabric as possible (e.g., shape of the folds, angle of swing). View-dependent appearance features for each fabric are also visible in this way. We ensured that the movement was as similar as possible for each fabric.

To create computer generated replicas of the reference fabrics, we needed to emulate both the appearance and the dynamics. Note that

140 appearance refers to the spatially varying reflected radiance of the  
 141 cloths, which depends on several factors such as the texture pat-  
 142 tern or the optical properties of the fabrics (e.g.: albedo or surface  
 143 scattering). All pieces of cloth were rendered using path tracing  
 144 with deferred shading [Eisenacher et al. 2013], simulating rough di-  
 145 electric materials with diffuse transmittance, together with albedo,  
 146 bump and opacity textures. For these, a set of close-up pictures  
 147 perpendicular to the fabrics was taken to generate tileable seam-  
 148 less textures representing patches of 30x30 cm. The only exception  
 149 was polyester satin; given its more anisotropic reflectance and color  
 150 shifts, we relied on the empirical microcylinder model of Sadeghi  
 151 and colleagues [2013]. Figure 2, shows the appearance of the final  
 152 CG replicas.

153 The dynamics of the different fabrics were simulated by model-  
 154 ing the cloth as a triangular mesh, along with proximity forces to  
 155 prevent primitives near each other from colliding, as proposed by  
 156 Baraff and Witkin [1998]. Similarly, we use additional constraints  
 157 for cloth-object collisions. If continuous time collisions remain af-  
 158 ter the initial solve, we rely on the robust collision algorithm from  
 159 Bridson et al. [2002], augmented by a fail-safe that cancels impact  
 160 while maintaining sliding motion [Harmon et al. 2008]. We re-  
 161 lied on physical parameters given by the manufacturer when avail-  
 162 able (such as density and thickness, e.g., burlap weighs  $207g\ m^2$   
 163 with 0.69mm thickness, while the values for silk are  $207g\ m^2$  and  
 164 0.69mm); all the remaining parameters were manually adjusted to  
 165 obtain a result as close as possible to the real cloth properties (see  
 166 Figure 4).

167 We then rendered all possible combinations of appearance and dy-  
 168 namics, yielding  $7 \times 7 = 49$  videos (six seconds each) replicating the  
 169 movement in the recorded video. Thus for each row (column) of  
 170 the matrix, only one rendered video matches the appearance with  
 171 the correct dynamics. In addition, to study the effect of viewing  
 172 distance on the perception of mismatched properties, we rendered  
 173 all of the stimuli at three different camera distances, resulting in  
 174 resolutions of  $1728 \times 1123$ ,  $1000 \times 650$  and  $520 \times 338$  from close to  
 175 far viewing distances respectively. A selection from this full set  
 176 of  $49 \times 3 = 147$  videos is included in this submission as supplemen-  
 177 tary material (the full set exceeds the upload limit). Note that we  
 178 rendered all videos with the swivel stool rotating in the opposite di-  
 179 rection from the real videos, to avoid that participants would base  
 180 their judgments on exact visual matching.

## 181 4 Experiments

182 To answer the questions set out in our introduction, we conducted  
 183 two perception experiments with 63 naive participants (34F/29M,  
 184 aged 18–27) with varying levels of experience in computer graph-  
 185 ics. We counterbalanced the order in which they performed Exper-  
 186 iment 1 and Experiment 2, to avoid ordering effects.

### 187 4.1 Experiment One: Ground Truth comparison

188 The goal of the first experiment is twofold: firstly, to evaluate how  
 189 effective the simulations were at capturing the appearance and dy-  
 190 namics of the real stimuli; and secondly, to determine whether ei-  
 191 ther dynamics or appearance were more important when animating  
 192 photo-realistic cloth.

193 We chose an experimental design where each participant only  
 194 watches a subset of the stimuli, in order to avoid fatigue effects.  
 195 Thus, the stimuli are distributed among participants ensuring that  
 196 each video is seen by 45 different people, and each person sees 105  
 197 different samples of the total set of 147.

198 Two equally calibrated screens of the same model were used for the



**Figure 4:** Comparison between the movements of the real cloth samples and the CG replicas. The first row shows the cotton rotating at the maximum speed. The second row shows the burlap at the frame just before starting to stabilize. Note that the real and CG samples are rotating in the same direction in these images just for comparison, but do so in opposite directions during the experiments to avoid exact image matching. To emulate the cloth motion, we paid special attention to the number, size and shape of the folds created (both at static and dynamic frames), the amount of bouncing, the effect of air forces, and the maximum height and width reached when rotating. For further comparisons, a selection of the videos are included in the supplementary material.

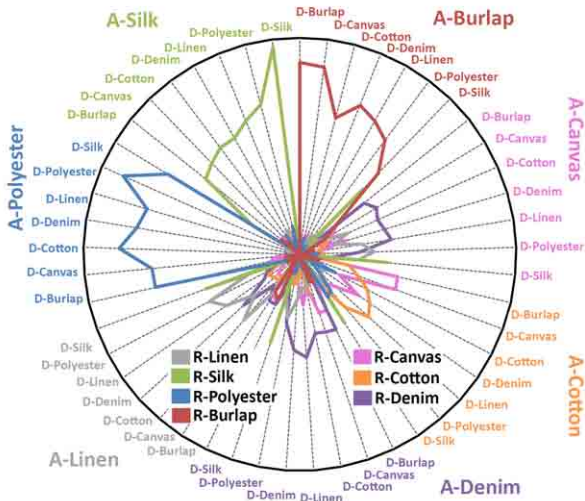


**Figure 5:** Two screen layout for the experiment 1. On the left, the navigation screen with the seven real (ground truth) reference fabrics. Each thumbnail has a radio button for selection and a replay button. On the right, the CG cloth that is currently being displayed.

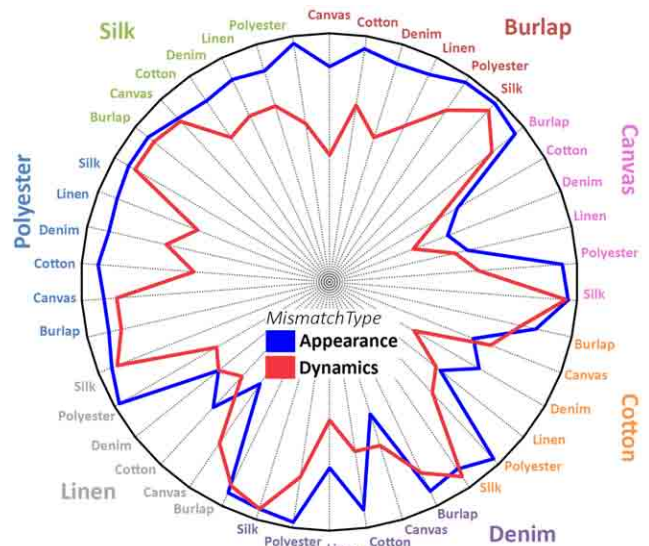
199 experiment (Dell U2311H IPS FullHD 23”). On the right screen,  
 200 one of the 147 rendered videos is shown, and the participant is  
 201 asked the question: ‘Which of the reference cloths on the left best  
 202 matches the one on the right?’. The participant can answer by  
 203 choosing any of the seven reference cloths shown in thumbnails  
 204 on the left (Figure 5). She can replay any of these reference ground  
 205 truth videos again, as many times as needed until an answer is given  
 206 (there is no time limit). Each time a reference video is replayed at  
 207 full resolution on the left, the current CG replica that is being eval-  
 208 uated is played on the right for comparison purposes. Both videos  
 209 are synchronized, but the cloths rotate in opposite directions to dis-  
 210 courage exact visual pattern matching.

211 At the start of the experiment, we ensure all participants have famil-  
 212 iarized themselves with all real stimuli. All participants are shown  
 213 a representative frame of every one of the seven reference videos as  
 214 a thumbnail on the left screen. They view all of the videos by click-  
 215 ing on each of these thumbnails, and the corresponding six-second  
 216 video is played on the right screen. They can repeat each one as  
 217 many times as needed. The experiment took between 25 and 45  
 218 minutes, separated in two halves by a 5-minute break.





**Figure 6:** Experiment 1 results, summarized as a radar graph and collapsed over distance (which had no effect). The colored areas in the graph represent how often each Response was given for the Appearance/Dynamics combinations depicted on the perimeter.



**Figure 7:** Experiment 2 results, summarized as a radar graph and collapsed over distance (which had no effect). The outermost labels on the perimeter indicate the “correct” fabric, while the innermost ones show the mis-matched one. The two line graphs indicate the percentage of mismatches accurately detected for the two types of mismatch: appearance or dynamics.

219 **Experiment One: Results.** Because of the way we designed our  
 220 experiment, we were able to cross-tabulate all participant responses  
 221 by summarizing them in a *Multi-way Frequency Table*. The variable  
 222 combinations for which frequency counts were calculated were: (1)  
 223 *Distance* x 3 (close, medium, far), (2) *Appearance* x 7 (denoted A-  
 224 Burlap, A-Canvas, A-Cotton, A-Denim, A-Linen, A-Polyester, A-  
 225 Silk), (3) *Dynamics* x 7 (D-Burlap – D-Silk) and (4) *Response* x 7  
 226 (R-Burlap – R-Silk). The results are shown in Figure 6.

227 We then analyzed these data using Log-Linear Analysis, which al-  
 228 lows us to find the best model to fit the observed data. In the case  
 229 of Figure 6, the best model was (2,4), (3,4), meaning that there was  
 230 a main effect of both Appearance(2) and Dynamics (3) on the Re-  
 231 sponse (4) given. However, the distance from the camera had no  
 232 effect on the responses. From Figure 6 we can see that appear-  
 233 ance dominated the responses for three fabrics: Burlap, Silk and  
 234 Polyester. There was more confusion between the other materials.  
 235 We also looked at how often Dynamics affected the choices, and  
 236 the only material where dynamics was very influential was for Silk,  
 237 where the green line in the figure shows how the response was al-  
 238 ways silk when the dynamics were silk, and silk was also often  
 239 picked when the appearance was a different material (e.g., see the  
 240 green spike for A-Burlap).

## 241 4.2 Experiment Two: Identifying Mismatches

242 The main goal of this experiment is to determine how accurate partic-  
 243 ipants were at identifying mismatches between the appearance and  
 244 dynamics of photo-realistic cloth animations. First, as in Ex-  
 245 periment One, participants are shown the seven real videos at the  
 246 beginning and are allowed to replay them until they become famil-  
 247 iar with them. Once the test begins, one of the recorded videos  
 248 is shown on the left screen while two CG videos from our stimu-  
 249 li matrix are shown side-by-side on the right screen. One of the  
 250 CG videos is always the corresponding replica of the real video  
 251 shown, with matching appearance and dynamics, while the other  
 252 one has been rendered with either the appearance or the dynam-  
 253 ics from a different cloth. The order is randomized for each pair

254 of stimuli. This leads to 252 combinations in total: 7 fabrics x  
 255 12 mismatched options (6 each for appearance and dynamics) x 3  
 256 viewing distances. The participant is asked which of the two simu-  
 257 lated cloths on the right is most similar to the ground-truth cloth  
 258 video shown on the left. There is no time limit, and the participant  
 259 is allowed to replay the videos as often as necessary.

260 As in the previous experiment, we opted for an experimental design  
 261 where each participant only watches a subset of the stimuli in order  
 262 to avoid fatigue effects. Thus, the stimuli are distributed so as to  
 263 ensure that each stimulus pair is seen by 45 different people, and  
 264 each person sees 180 different samples of the total set of 252.  
 265 This experiment lasted between 50 and 70 minutes, again divided  
 266 in two parts by a break of 5–10 minutes. The experiment was per-  
 267 formed using the same screens and controlled settings as in Exper-  
 268 iment One.

269 **Experiment Two: Results.** As in the previous experiment, we  
 270 were able to cross-tabulate all participant data by summarizing  
 271 the percentage of correctly identified mismatches in a multi-way  
 272 frequency table, and statistically analyzed them using Log-Linear  
 273 Analysis. Again, distance had no effect on the results, but both  
 274 Appearance, Dynamics, and their interaction did. The results are  
 275 shown in Figure 7. We can again see that appearance mismatches  
 276 were most easily detected for most, but not all, fabrics, whereas  
 277 participants were more confused about the dynamics mismatches.

## 278 5 Conclusions

279 In this paper, we have presented the results of two perceptual exper-  
 280 iments where we explored the interactions of appearance and dyn-  
 281 amics of seven common woven fabrics. We demonstrate how ap-  
 282 pearance dominates over dynamics, except for the few cases where  
 283 dynamics are very characteristic, such as in the case of silk. We  
 284 also found that these effects are robust across different viewing dis-  
 285 tances.

286 As future work, it would be interesting to consider some other factors that may have an effect on the perception of moving cloth (e.g. different illumination conditions such as environment lighting), or to explore more deeply the influence of the most important factors of cloth simulation considered here (e.g. BRDF and spatial frequency of the textures in the case of the appearance, dynamics parameters in the case of motion synthesis). Finally, performing a similar study with animated characters wearing clothes made from these fabrics would allow us to confirm our findings in more ecologically valid and familiar scenarios.

## 296 Acknowledgements

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