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Automated outdoor depth-map generation and alignment^{*}

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ABSTRACT

Image enhancement tasks can highly benefit from depth information, but the direct estimation of outdoor depth maps is difficult due to vast object distances. This paper presents a fully automatic framework for model-based generation of outdoor depth maps and its applications to image enhancements. We leverage 3D terrain models and camera pose estimation techniques to render approximate depth maps without resorting to manual alignment. Potential local misalignments, resulting from insufficient model details and rough registrations, are eliminated with our novel free-form warping. We first align synthetic depth edges with photo edges using the as-rigid-as-possible image registration and further refine the shape of the edges using the tight trimap-based alpha matting. The resulting synthetic depth maps are accurate, calibrated in the absolute distance. We demonstrate their benefit in image enhancement techniques including reblurring, depth-of-field simulation, haze removal, and guided texture synthesis.

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1 1. Introduction

A limited configuration in taking photographs does not al-2 3 ways lead to the highest quality, and often motivates enhancement of photographs. Computational photography has addressed 4 5 such limitations, which introduces additional flexibility on focus, exposure, and depth [1-3]. Among them, depth information, on 6 which we focus here, can greatly facilitate diverse image manipu-7 lations, such as refocusing, dehazing, texture synthesis, and image 8 editing [4–7]. 9

Outdoor photographs (e.g., natural landscapes) represent by far the biggest group in many media services [8], but their direct depth estimation poses a challenge. They are usually monocular, which has a low chance to work with typical structure-frommotion. Within-image features, such as airlights or textures [9,10], help, but are not always available. Range sensors [11] are applicable only to limited distance ranges (only up to tens of meters).

One better alternative can be an indirect estimation from a 3D terrain model, which renders the reference depth map as previously suggested by Kopf et al. [12]. The terrain model is already available for the whole planet and recent photographs are usually

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https://doi.org/10.1016/j.cag.2018.05.001 0097-8493/© 2018 Elsevier Ltd. All rights reserved. geo-tagged (e.g., the global positioning system; GPS), which can21serve as a strong external cue. Further, this approach can be dis-22tinguished for its higher resolution and accuracy; the direct esti-23mation may yield a coarser resolution or wrong outcomes.24

However, the 3D terrain model may be insufficient in its resolution and details (e.g., textures and objects). Also, a precise registration between real and virtual views is challenging, where the alignment errors result in visible artifacts in edited images. Manual registration starting at a rough initial guess can help [13], but is laborious and inappropriate for massive batch processing. 31

In this work, we present a fully automatic framework for depth-32 map generation and alignment for an outdoor photograph. A vir-33 tual camera is first localized with the geo-tagging information of 34 a photo and recent camera pose estimation techniques. Then, the 35 3D terrain model is rendered at the virtual camera to produce an 36 initial approximate depth map. Local inaccuracies, resulting from 37 the rough registration and insufficient details of the model, are 38 subsequently reduced using our novel automatic free-form warp-39 ing. We first align discontinuities in the synthetic depths with 40 photo edges using the as-rigid-as-possible image registration. The 41 shape of the edges is further refined using the tight trimap-based 42 alpha matting. The resulting depth map, synthesized from the 43 geo-referenced terrain model, is absolute (calibrated in meters). 44 We show its benefit in image enhancement tasks, including re-45 focusing, defocus manipulation (see Fig. 1), dehazing, and texture 46 synthesis. 47

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Fig. 1. Transforming an outdoor photograph into a model-like look. An automatically generated synthetic depth map is used to calculate plausible blur kernel size map (middle) to simulate shallow depth-of-field (right) in landscape images (left), where such an effect cannot be achieved using standard optics for physical reasons. Virtual camera: full-frame, f-number=1.0, focal length=1200 mm, focus distance=5 km.

2.3.

48 2. Related work

We briefly review previous work on depth map reconstruction
and its major applications including defocus manipulation and de hazing.

52 2.1. Depth map reconstruction

Robust depth map reconstruction is an ongoing subject of interest. A typical approach is to rely on stereo image pairs [18,19] or multiple/multiview images [20–22]. More recently, short-distance range-sensing devices, such as Kinect [11], improved the availability of depth maps in indoor environments [23].

Multiple images are not easily available in practice, and 58 single-image processing has also been intensively studied. Semi-59 60 automatic user interaction often helps [24–26], and further modification to hardware or light patterns proved its utility, such as light 61 62 fields [27–31], coded aperture [32] or structured light method [33]. Depth maps can further be generated semi-automatically using 63 sparse samples (seeds) provided by the user [34–37]. In these ap-64 proaches, anisotropic diffusion is used to propagate depth informa-65 66 tion from seeds to the rest of the image. The key assumption here

is that the gradient of the resulting depth map should roughly cor-respond to the color gradient.

The previous methods are not applicable to ours which is tar-69 70 geting on single outdoor photographs. The stereo vision techniques 71 require multiple images, while the range sensors work only for a small range of distances. Computational photography requires spe-72 cial hardware or modification to the aperture. For the diffusion-73 based techniques, real photos often violate their main assumption 74 75 about compatible depth and color gradients, and also, the positions 76 of depth seeds require to be accurate. Otherwise, the diffusion will 77 propagate small misalignments to a notably larger area and lead to 78 notable artifacts.

Kopf et al. showed the combination of geo-tagging and 3D mod-79 80 els can be used for fairly accurate geo-registration and many ap-81 plications including dehazing and relighting [12]. The geo-tagging already allows us to select an effective subset for structure from 82 motion [38], but when combining with the available 3D models, 83 we can directly render a depth map. The combination even enables 84 to assign geo-locations and labels onto pixels [39], point clouds 85 86 [38,40], or annotate photos [41]. Nevertheless, the depth map is not pixel-perfect and requires fine alignment; we address this is-87 88 sue in the present work.

While the majority of previous approaches estimate *relative* depth maps, our solution can generate *absolute* depth map from the geo-referenced digital terrain model. This is highly beneficial in many image enhancement applications; for instance, there are more chances in estimating kernels for defocus blur or other effects. Similarly to our goals, the method proposed by Kopf et al. 95 [12] allows to synthesize absolute depth maps. However, it requires a *user-assisted* interaction for registration, which motivates for our novel depth free-form warping step (Section 3). 98

2.2. Defocus manipulation

The defocus blur, caused by shallow depth of field (DOF), is pronounced in indoor photos or films, but hardly exhibited in outdoor photography (even with large lenses). Its capture is inherently restricted to a particular configuration (e.g., focus and *f*-number). Thus, its post-reproduction for novel configurations (with computational photography) drew attention for refocusing [32,33] or defocus magnification [42].

Another mainstream is the postprocessing of a usual single-107 view image, which comprises deblur and refocusing. Typically, the 108 deblur involves blind deconvolution using known priors to esti-109 mate kernels [24,43]. However, a precise solution requires to con-110 sider geometric visibilities similarly to the distributed ray tracing 111 [44]. Our solution and its absolute depth information can facilitate 112 the estimation of per-pixel local blur kernels to some extents, en-113 abling non-blind deconvolution. Refocusing can also benefit from 114 ours, which can use a precise rendering technique [45]. 115

Outdoor photographs are often hazed by atmospheric scattering 117 that can be characterized by medium transmission maps, depend-118 ing heavily on depths. Most previous work has focused on depth 119 estimation and radiance recovery, including Markov random fields 120 (MRF) [46], independent component analysis [47], dark channel 121 prior [48], factorial MRF [49–52], and machine-learning approaches 122 using random forests [53,54], and convolutional neural networks 123 [55]. In our case, geo-referenced absolute depth maps are used 124 for dehazing, which can be more precise than the previous ones. 125 We obtain such maps automatically without the previous manual 126 image-to-model registration [12]. 127

3. Automatic depth-map generation and alignment

In this section, we describe our fully-automatic approach to 129 depth-map generation from a single color image as well as a technique used for the final depth map refinement (see Figs. 2 and 3 for summary). 132

Retrieval of 3D terrain model. A Google-Earth-like digital terrain 133 models are currently available for the whole planet. They are acquired from satellites and/or planes and published in form of georeferred digital elevation maps (DEMs) even for less accessible regions. Such models are sufficient for our purposes (i.e., outdoor photographs), however in general case, a 3D model might not be

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