Learning a Discriminative Prior for Blind Image Deblurring

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$$B = I \bigotimes k + n$$

ill-posed, need additional prior knowledge.

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 $p(blurred\ image|clear\ image) \approx 0$

 $p(blurred\ image|blurred\ image) \approx 1$

Hand-craft features are limited and cannot generalize to real cases.

Learn this prior by a discriminator f(I) . We say $f(I) = p(blured\ image|I).$ Cross entropy.

Done by optimization.

$$\min_{I,k} \|I \bigotimes k - B\|_2^2 + \gamma \|k\|_2^2 + \mu \|\nabla I\|_0 + \lambda f(I)$$

I-sub problem

$$\min_{I}\|I\bigotimes k-B\|_2^2+\mu\|\nabla I\|_0+\lambda f(I)$$

Both f(I) and $\|\nabla I\|_0$ are highly non-convex.

Half-quadratic splitting trick, introduce two variables u and g and say $u \approx I, \, g \approx \nabla I.$

$$\min_{I,g,u} \|I \bigotimes k - B\|_2^2 + \alpha \|\nabla I - g\|_2^2 + \beta \|I - u\|_2^2 + \mu \|g\|_0 + \lambda f(u)$$

I-sub part.

$$\min_{I}\|I\bigotimes k-B\|_2^2+\alpha\|\nabla I-g\|_2^2+\beta\|I-u\|_2^2$$

Which has a closed-form solution

$$I = F^{-1}(\frac{\overline{F(k)}F(B) + \beta F(u) + \alpha(\sum_{t \in \{h,v\}} \overline{F(\nabla t)}F(g_t))}{\overline{F(k)}F(k) + \beta + \alpha \overline{F(\nabla t)}F(\nabla_t))})$$

$$\min_g \alpha \|\nabla I - g\|_2^2 + \mu \|g\|_0$$

Pixel-wise optimization problem.

$$g = [\|\nabla I\|^2 \ge \frac{\mu}{\alpha}]\nabla x$$

I-sub problem

u-sub part

$$\min_{u} \beta \|I - u\|_2^2 + \lambda f(u)$$

By gradient descent.

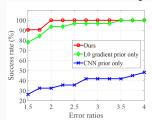
k-sub problem

k-sub problem

$$\min_{k} \|\nabla I \bigotimes k - \nabla B\|_2^2 + \gamma \|k\|_2^2$$

which can also be efficiently solved by FFT (?).

Effectiveness of CNN prior.





Authors



With adversarial process, the discriminator will be more effective.

The gradient ∇f is the quickest way to change the value of f(I) but not the realness of I.

Just use discriminator trick to implicitly learn the prior knowledge $p(blurred\ image|I).$

And this trick is of little effectiveness.

Conclusion