Result Table:1

Image	Sigma=20	Sigma=30
Peppers	Bilateral	24.6703
	Noised image	18.7662
Barbara	Bilateral	24.1778
	Noised image	18.7937
Lena	Bilateral	27.7501
	Noised image	18.7019

V. CONCLUSION

In this paper I have designed a scientific analysis of the optimal parameters value for bilateral filter in image denoising and present a multi resolution image framework which is combined with bilateral filter and wavelet threshold. Here I proposed decompose an image in low pass and high pass frequency components and apply bilateral filter method on the approximation sub bands. We got the maximum $\sigma/_r$ value of bilateral filter is linearly related to the standard derivation of noise. The maximum value of the $\sigma/_d$ is relatively separate of the noise power. Outcomes of the result we calculate the noise standard derivation for every level of the sub band decomposition and use a continues several of it for the $\sigma/_r$ value for bilateral filter.

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