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CLASS	SUBJECT <u>IPCV</u>	
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Bilateral Filtering Pseudocode.

1. Define Input.

Image (I)

Kernel Size (d) - diameter of the neighborhood around the center pixel.

Sigma- x (σ_x) - controls the influence of intensity similarity.

(σ_s) - controls the influence of spatial proximity.

2. Iterate over each pixel (i, j) in the image

Initialize output pixel intensity ($I_{\text{result}}[i, j]$) to 0

Initialize normalization factor (W) to 0.

3. Iterate over neighbouring pixel (p, q) within the Kernel window centered at (i, j).

Calculate spatial distance (σ_s) between (i, j) & (p, q).

Calculate intensity difference (σ_x) between (i, j) & (p, q).

Calculate spatial weight (w_s) using a Gaussian funcⁿ with standard deviation σ_s : $w_s = \exp(-\sigma_s^2 / (2 * \sigma_s^2))$

Calculate range weight (w_r) using a Gaussian function with standard deviation σ_x : $w_r = \exp(-\sigma_x^2 / (2 * \sigma_x^2))$

Combine weights: $w = w_s * w_r$.

Update Normalizer factor: $W += w$

Accumulate weighted intensity: $I_{\text{result}}[i, j] += I(p, q) * w$

4. Normalize Output.

$I_{\text{result}}[i,j] = W$ (divide by the normalizer factor).

Experimenting with σ_r & σ_s (σ_s)

1. Implement the bilateral filtering function with options to set σ_r and σ_s .

2. Run the function on your image with different combinations of σ_r and σ_s values (eg: low, medium, high values for each).

3. Observe the resulting images. Typically:

- Lower σ_r leads to sharper edge preservation but may not remove all noise.
- Higher σ_r removes more noise but can blur edges.
- Lower σ_s affects a smaller neighborhood potentially leading to patchy results.
- Higher σ_s considers a larger neighborhood, provide smoother filtering.

Comparing with OpenCV:

- Use OpenCV's `cvtColor_BilateralFilter` function on your image, with similar settings for kernel size, σ_r & σ_s as your implementation.
- Calculate the absolute difference between the intensity values of corresponding pixels in your result (I_{result}) & open CV result (I_{input}).

- Sum the absolute differences across all pixels to get the total error (abs. sum $(I_{\text{inbuilt}} - I_{\text{result}})$)

Conclusion.

Analyzing the visual quality of your bilateral filtering results with different parameter settings.

Compare the error between your implementation & OpenCV's function. A small error indicates similar filtering behaviour.

Consider the tradeoff between noise removal & edge preservation based on your image & application requirement.