## Supplementary Table S 3

- 1. Color Histogram [17]
- 2. Multiscale Histograms [20]
- 3. Haralick Textures (Hue()) [9]
- 4. Pixel Intensity Statistics (Chebyshev (Hue())) [2]
- 5. Multiscale Histograms (Chebyshev (Fourier ())) [9]
- 6. Zernike Coefficients (Wavelet ()) [19]
- 7. Color Histogram [4]
- 8. Zernike Coefficients (Fourier ()) [22]
- 9. Pixel Intensity Statistics [4]
- 10. Chebyshev Coefficients (Fourier (Edge ())) [8]
- 11. Color Histogram [5]
- 12. Chebyshev-Fourier Coefficients (Fourier (Hue())) [12]
- 13. Gabor Textures [4]
- 14. Pixel Intensity Statistics (Edge ()) [4]
- 15. Zernike Coefficients (Fourier ( Edge ())) [9]
- 16. Zernike Coefficients (Fourier (Edge ())) [17]
- 17. Pixel Intensity Statistics [3]
- 18. Comb Moments (Hues ()) [44]
- 19. Color Histograms [3]
- 20. Comb Moments (Wavelet (Fourier ())) [9]

## Table S 3. Top 20 features from mRMR feature selection.

Top 20 features in the training dataset identified by mRMR feature selection. A complete list of

features computed by WND-CHARM can be found in Orlov et. al. 2008 (8).

## SUPPLEMENTAL REFERENCES

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