

ELECTRONIC SUPPLEMENTARY MATERIAL

The effect of preprocessing filters on predictive performance in radiomics

Full list of extracted features

Common features for all preprocessing filters

firstorder_10Percentile

firstorder_90Percentile

firstorder_Energy

firstorder_Entropy

firstorder_InterquartileRange

firstorder_Kurtosis

firstorder_Maximum

firstorder_Mean

firstorder_MeanAbsoluteDeviation

firstorder_Median

firstorder_Minimum

firstorder_Range

firstorder_RobustMeanAbsoluteDeviation

firstorder_RootMeanSquared

firstorder_Skewness

firstorder_TotalEnergy

firstorder_Uniformity

firstorder_Variance

glcm_Autocorrelation

glcm_ClusterProminence
glcm_ClusterShade
glcm_ClusterTendency
glcm_Contrast
glcm_Correlation
glcm_DifferenceAverage
glcm_DifferenceEntropy
glcm_DifferenceVariance
glcm_Id
glcm_Idm
glcm_Idmn
glcm_Idn
glcm_Imc1
glcm_Imc2
glcm_InverseVariance
glcm_JointAverage
glcm_JointEnergy
glcm_JointEntropy
glcm_MaximumProbability
glcm_SumEntropy
glcm_SumSquares
gldm_DependenceEntropy
gldm_DependenceNonUniformity
gldm_DependenceNonUniformityNormalized
gldm_DependenceVariance
gldm_GrayLevelNonUniformity

glbm_GrayLevelVariance
glbm_HighGrayLevelEmphasis
glbm_LargeDependenceEmphasis
glbm_LargeDependenceHighGrayLevelEmphasis
glbm_LargeDependenceLowGrayLevelEmphasis
glbm_LowGrayLevelEmphasis
glbm_SmallDependenceEmphasis
glbm_SmallDependenceHighGrayLevelEmphasis
glbm_SmallDependenceLowGrayLevelEmphasis
gIrlm_GrayLevelNonUniformity
gIrlm_GrayLevelNonUniformityNormalized
gIrlm_GrayLevelVariance
gIrlm_HighGrayLevelRunEmphasis
gIrlm_LongRunEmphasis
gIrlm_LongRunHighGrayLevelEmphasis
gIrlm_LongRunLowGrayLevelEmphasis
gIrlm_LowGrayLevelRunEmphasis
gIrlm_RunEntropy
gIrlm_RunLengthNonUniformity
gIrlm_RunLengthNonUniformityNormalized
gIrlm_RunPercentage
gIrlm_RunVariance
gIrlm_ShortRunEmphasis
gIrlm_ShortRunHighGrayLevelEmphasis
gIrlm_ShortRunLowGrayLevelEmphasis
glszm_GrayLevelNonUniformity

glszm_GrayLevelNonUniformityNormalized
glszm_GrayLevelVariance
glszm_HighGrayLevelZoneEmphasis
glszm_LargeAreaEmphasis
glszm_LargeAreaHighGrayLevelEmphasis
glszm_LargeAreaLowGrayLevelEmphasis
glszm_LowGrayLevelZoneEmphasis
glszm_SizeZoneNonUniformity
glszm_SizeZoneNonUniformityNormalized
glszm_SmallAreaEmphasis
glszm_SmallAreaHighGrayLevelEmphasis
glszm_SmallAreaLowGrayLevelEmphasis
glszm_ZoneEntropy
glszm_ZonePercentage
glszm_ZoneVariance
ngtdm_Busyness
ngtdm_Coarseness
ngtdm_Complexity
ngtdm_Contrast
ngtdm_Strength

Features in Original

shape_Elongation
shape_Flatness
shape_LeastAxisLength
shape_MajorAxisLength

shape_Maximum2DDiameterColumn
 shape_Maximum2DDiameterRow
 shape_Maximum2DDiameterSlice
 shape_Maximum3DDiameter
 shape_MeshVolume
 shape_MinorAxisLength
 shape_Sphericity
 shape_SurfaceArea
 shape_SurfaceVolumeRatio
 shape_VoxelVolume

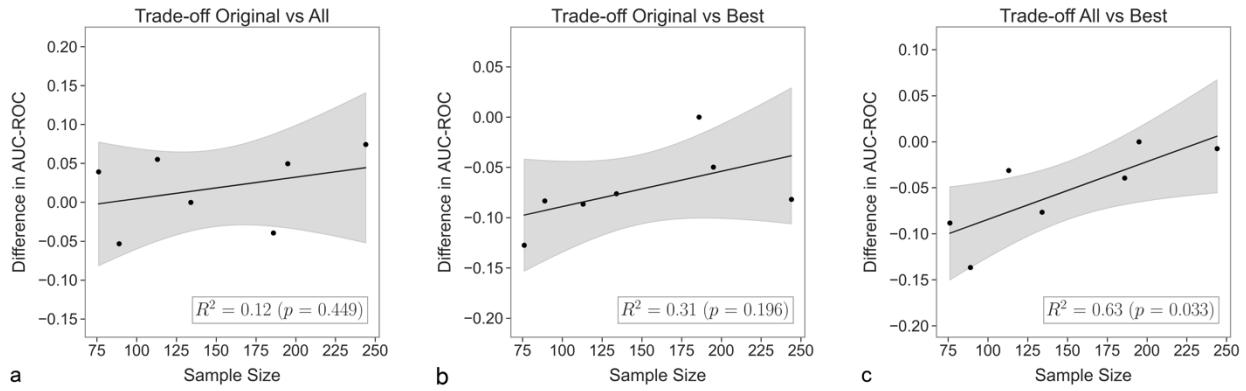
Results for quantization with a fixed number of bins

Using a fixed number of bins ($N = 100$) instead of a fixed bin width ($N = 25$) for quantization, the following results were obtained:

	AUC-ROC Original	AUC-ROC All	AUC-ROC Tuned	$\Delta_{\text{All - Original}}$	$p_{\text{All - Original}}$	$\Delta_{\text{Tuned - Original}}$	$p_{\text{Tuned - Original}}$	$\Delta_{\text{Tuned - All}}$	$p_{\text{Tuned - All}}$	Best Filter
CRLM	0.53 (0.4-0.66)	0.57 (0.44-0.7)	0.66 (0.53-0.78)	0.04	0.665	0.13	0.06	0.09	0.288	lbp-3D
Desmoid	0.73 (0.66-0.81)	0.78 (0.71-0.86)	0.78 (0.71-0.85)	0.05	0.247	0.05	0.224	0	0.865	log-sigma-
GIST	0.68 (0.61-0.75)	0.76 (0.69-0.82)	0.76 (0.7-0.82)	0.08	0.01	0.08	0.001	0	0.781	square_-
HN	0.78 (0.7-0.86)	0.78 (0.7-0.86)	0.85 (0.79-0.92)	0	0.996	0.07	0.028	0.07	0.053	square_-
Lipo	0.77 (0.68-0.85)	0.82 (0.74-0.9)	0.85 (0.78-0.92)	0.05	0.281	0.08	0.035	0.03	0.504	square_-
Liver	0.71 (0.63-0.78)	0.67 (0.59-0.75)	0.71 (0.63-0.78)	-0.04	0.366	0	1	0.04	0.366	original_-
Melanoma	0.54 (0.42-0.67)	0.49 (0.37-0.61)	0.63 (0.51-0.75)	-0.05	0.526	0.09	0.24	0.14	0.088	square_-

Although there are some differences visible when comparing with Table 4 (e.g., in AUC-ROC Original for CRLM and Lipo), the main conclusions still hold: In no case the AUC-ROCs of the original feature set are significantly better than those of the set of all features. Also, tuning the preprocessing filter never significantly loses performance.

Similarly, there is no large change in the trade-offs and the correlations:



	Original	Exponential	Gradient	LBP	LoG	Logarithm	Square	Squareroot	Wavelet
Original	1.0	0.759	0.801	0.665	0.765	0.714	0.8	0.842	0.838
Exponential	0.759	1.0	0.641	0.574	0.643	0.581	0.708	0.674	0.691
Gradient	0.801	0.641	1.0	0.684	0.813	0.644	0.708	0.73	0.807
LBP	0.665	0.574	0.684	1.0	0.674	0.6	0.596	0.634	0.717
LoG	0.765	0.643	0.813	0.674	1.0	0.618	0.698	0.693	0.777
Logarithm	0.714	0.581	0.644	0.6	0.618	1.0	0.647	0.906	0.672
Square	0.8	0.708	0.708	0.596	0.698	0.647	1.0	0.694	0.728
Squareroot	0.842	0.674	0.73	0.634	0.693	0.906	0.694	1.0	0.758
Wavelet	0.838	0.691	0.807	0.717	0.777	0.672	0.728	0.758	1.0