

Supplementary Materials: Impact of Interobserver Variability in Manual Segmentation of Non-Small Cell Lung Cancer (NSCLC) Applying Low-Rank Radiomic Representation on Computed Tomography

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Supplementary Materials: The following are available online at www.mdpi.com/xxx/s1, Supplement Table S1: The CT parameters for Stanford NSCLC Radiogenomics dataset, Supplement Table S2 The CT parameters for Harvard NSCLC Radiomics-Genomics dataset, Supplement Table S3: Similarity of the radiomic signatures using Pearson correlation among different segmenters are presented for different stratifications based on CT parameters, Supplement Table S4: Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter while there is Contrast-Enhancement (CE), Supplement Table S5: Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter while there is Non-Contrast-Enhanced (UN), Supplement Table S6: Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter for higher convolutional kernel (CKh). Supplement Table S7: Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter for slice thickness between 2mm and 4mm. Table S8. Itraclass correlation coefficient based on radiomics categories and with the respect of different group means. For each segmenter, mean and standard deviation of correlation coefficient is calculated for every radiomics' category, Table S9. Radiomic features with lesser stability with the respect to different segmenters. Means and standard deviations of these radiomics are presented. Table S10. More detailed information about the Radiomic features used in this study., Table S11. The hazard ratio for each covariate in the maximal cox proportional hazard model.

Table S1. The CT parameters for Stanford NSCLC Radiogenomics dataset.

CT Imaging Parameters				
Characteristic	Siemens Healthineers	Ge Healthcare	Philips Medical Systems Tech.	Toshiba Corporation
Contrast Enhanced	24	71	2	-
Non- Contrast Enhanced	34	71	4	1
Convolution Kernel				
Standard (soft-tissue)		43		-
Hard (lung, bone, boneplus)	4 54	99	3 (2(YB), 1(B)) 3(C)	- 1 (FC52)
Exposure (mAs)	159.5 (30, 466)	447.5 (31.8,754)	217 (93, 352)	217
Focal Spots (mm)				
0.8	1	14	-	-
1.2	57	118	-	1
Unknown	-	10	6	-
Tube Voltage (kVp)	120 (100,130)	120 (100,140)	unknown	unknown
Reconstruction Diameter (mm)	358.5 (235,500)	400 (302,500)	350 (304,384)	474.6
Slice Thickness (mm)				
≤1	49	14	3	-
1< .≤2	4	100	-	1
2< ≤4	4	27	2	-
≤5	1	1	1	-

Table S2. The CT parameters for Harvard NSCLC Radiomics-Genomics dataset.

CT imaging parameters			
Characteristic	Siemens Healthineers	Ge Healthcare	Philips Medical Systems Tech.

Contrast Enhanced	26	1	25
Non- Contrast Enhanced	12	-	23
Convolution Kernel			
Standard (soft-tissue)	3	-	40 (30:B,10:A)
Hard (lung, bone,boneplus)	35	1	8 (1:C, 7:D)
			-
Exposure (mAs)	125 (58, 158)	5	119 (20, 390)
Focal Spots (mm)			
0.95	6	1	-
1.2	32	-	-
Unknown	-	-	48
Tube Voltage (kVp)	120 (120,140)	120	-
Reconstruction Diameter (mm)	465 (308, 700)	402	430 (311, 600)
Slice Thickness (mm)			
≤ 1	-	-	-
$1 < . \leq 2$	-	-	4
$2 < \leq 4$	7	1	17
≤ 5	31	-	27

Table S3. Similarity of the radiomic signatures using Pearson correlation among different segmenters are presented for different stratifications based on CT.

NSCLC Dataset	Segmenters ID	Similarity Among Segmentations			
		Correlation Score			
		CE	UN	CKh	ST
LUNG3	BY	0.95	0.79	0.84	0.96
NSCLC-Radiomics-Genomics	LS	0.92	0.93	0.96	0.97
Harvard Dataset	MH	0.89	0.97	0.93	0.97
NSCLC-Radiogenomics	BY	0.92	0.96	0.87	0.91
Stanford Dataset	LS	0.86	0.97	0.78	0.91
	MH	0.74	0.89	0.53	0.90

CE-Contrast-enhanced; UN-unenhanced; CKh-convolutional kernel (hard); ST- slice thickness parameters.

Table S4. Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter while there is Contrast-Enhancement (CE).

NSCLC Datasets	Modeling covariates	Prediction Survival							
		BY		LS		MH		SK-GT	
		c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹
LUNG3	clinical & demographic ²							0.71	0.02
NSCLC-Radiomics-Genomics	Three PC radiomic signatures	0.66	0.06	0.67	0.04	0.63	0.3	0.7	0.03
Harvard Dataset	Radiomic signatures, clinical & demographic	0.73	0.2	0.71	0.2	0.7	0.3	0.72	0.2
NSCLC-Radiogenomics	clinical & demographic ³							0.71	0.02

Stanford Dataset CE	Three PC radiomic signatures	0.69	0.03	0.65	0.02	0.61	0.3	0.68	0.01
	Radiomic signatures, clinical & demographic	0.77	0.002	0.75	0.001	0.75	0.005	0.76	0.001

CI: confidence interval. 1. p value by likelihood ratio test versus the hypothesis that the model is no better than the null model. 2. Clinical and demographic covariates for LUNG3-NSCLC-Radiomics-Genomics Harvard Dataset: sex, stage status, and histology. 3. Clinical and demographic covariates for NSCLC-Radiogenomics Stanford Dataset: sex, stage status, and histology.

Table S5. Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter while there is Non-Contrast-Enhanced (UN).

		Prediction survival							
NSCLC Datasets	Modeling covariates	BY		LS		MH		SK-GT	
		c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹
LUNG3 NSCLC-Radiomics-Genomics Harvard Dataset	clinical & demographic ²							0.71	0.02
	Three PC radiomic signatures	0.69	0.03	0.65	0.02	0.61	0.3	0.68	0.01
UN	Radiomic signatures, clinical & demographic	0.77	0.002	0.75	0.001	0.75	0.005	0.76	0.001
NSCLC-Radiogenomics Stanford Dataset	clinical & demographic ³							0.66	0.08
	Three PC radiomic signatures	0.63	0.05	0.67	0.05	0.67	0.08	0.6	0.2
	Radiomic signatures, clinical & demographic	0.64	0.2	0.68	0.05	0.69	0.3	0.65	0.3

CI: confidence interval. 1. p value by likelihood ratio test versus the hypothesis that the model is no better than the null model. 2. Clinical and demographic covariates for LUNG3-NSCLC-Radiomics-Genomics Harvard Dataset: sex, stage status, and histology. 3. Clinical and demographic covariates for NSCLC-Radiogenomics Stanford Dataset: sex, stage status, and histology.

Table S6. Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter for higher convolutional kernel (CKh).

		Prediction survival							
NSCLC Datasets	Modeling covariates	BY		LS		MH		SK-GT	
		c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹
LUNG3	clinical & demographic ²							0.57	0.5

NSCLC-Ra- diomics-Ge- nomics	Three PC ra- diomic signa- tures	0.62	0.4	0.61	0.4	0.6	0.5	0.7	0.03
Harvard Da- taset CKh	Radiomic signatures, clinical & de- mographic	0.65	0.4	0.61	0.6	0.64	0.3	0.77	0.02
	clinical & de- mographic ³							0.7	<0.005
NSCLC-Ra- diogenomics	Three PC ra- diomic signa- tures	0.68	<0.005	0.7	0.003	0.67	0.04	0.7	<0.005
Stanford Da- taset CKh	Radiomic signatures, clinical & de- mographic	0.74	<0.005	0.76	<0.005	0.75	0.002	0.75	<0.005

CI: confidence interval. 1. p value by likelihood ratio test versus the hypothesis that the model is no better than the null model. 2. Clinical and demographic covariates for LUNG3-NSCLC-Radiomics-Genomics Harvard Dataset: sex, stage status, and histology. 3. Clinical and demographic covariates for NSCLC-Radiogenomics Stanford Dataset: sex, stage status, and histology.

Table S7. Overall survival, Cox regression. Using the low-rank representation of the radiomic signatures survival prediction is measured for each segmenter for slice thickness between 2mm and 4mm.

NSCLC Da- taset	Modeling covariates	Prediction survival							
		BY		LS		MH		SK-GT	
		c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹	c-statistic (95% CI)	p versus null ¹
	clinical & demo- graphic ²							0.7	0.4
NSCLC-Ra- diomics-Ge- nomics	Three PC ra- diomic signa- tures	0.69	0.6	0.61	0.8	0.67	0.4	0.64	0.5
Harvard Da- taset ST	Radiomic signatures, clinical & demo- graphic	0.77	0.5	0.72	0.7	0.78	0.2	0.77	0.4
	clinical & demo- graphic ³							0.66	0.5
NSCLC-Ra- dioge- nomics	Three PC ra- diomic sig- natures	0.5	0.9	0.63	0.7	0.58	0.9	0.6	0.7
Stanford Dataset ST	Radiomic signatures, clinical & demo- graphic	0.76	0.5	0.65	0.8	0.73	0.7	0.79	0.5

CI: confidence interval. 1. p value by likelihood ratio test versus the hypothesis that the model is no better than the null model. 2. Clinical and demographic covariates for LUNG3-NSCLC-Radiomics-Genomics Harvard Dataset: sex, stage status, and histology. 3. Clinical and demographic covariates for NSCLC-Radiogenomics Stanford Dataset: sex, stage status, and histology.

Table S8. Itraiclass correlation coefficient based on radiomics categories and with the respect of different group means. For each segmenter, mean and standard deviation of correlation coefficient is calculated for every radiomics' category.

	Similarity Among Segmenters for Radiomics Categories			
	Radiomics Category	Correlation BY	Correlation LS	Correlation MH
LUNG3 NSCLC-Radi-omics-Genomics Harvard Dataset	Shape-Based	0.75(± 0.15)	0.79(± 0.2)	0.82(± 0.14)
	1 st -Order	0.81(± 0.24)	0.85(± 0.2)	0.84(± 0.18)
	GLCM	0.71(± 0.27)	0.82(± 0.1)	0.74(± 0.21)
	GLSM	0.72(± 0.2)	0.7(± 0.2)	0.73(± 0.2)
	GLRLM	0.72(± 0.3)	0.79(± 0.2)	0.76(± 0.2)
	NGTDM	0.65(± 0.2)	0.71(± 0.16)	0.75(± 0.1)
	GLDM	0.73(± 0.3)	0.78(± 0.2)	0.77(± 0.2)
	LOG	0.78(± 0.1)	0.8(± 0.1)	0.81(± 0.2)
	Wavelet	0.88(± 0.14)	0.86(± 0.2)	0.89(± 0.17)
NSCLC-Radioge-nomics Stanford Dataset	Shape-Based	0.74(± 0.13)	0.72(± 0.14)	0.74(± 0.1)
	1st-Order	0.69(± 0.2)	0.67(± 0.16)	0.7(± 0.18)
	GLCM	0.73(± 0.2)	0.67(± 0.1)	0.7(± 0.16)
	GLSM	0.82(± 0.1)	0.7(± 0.1)	0.79(± 0.1)
	GLRLM	0.82(± 0.07)	0.8(± 0.1)	0.7(± 0.06)
	NGTDM	0.4(± 0.2)	0.48(± 0.2)	0.53(± 0.3)
	GLDM	0.8(± 0.1)	0.77(± 0.2)	0.76(± 0.1)
	LOG	0.82(± 0.1)	0.75(± 0.1)	0.77(± 0.2)
	Wavelet	0.83(± 0.2)	0.8(± 0.1)	0.78(± 0.2)

Table S9. Radiomic features with lesser stability with the respect to different segmenters. Means and standard deviations of these radiomics are presented.

Less Stable Radiomics		
Radiomic Types	Radiomics name	Correlation average
Shape-based	Sphericity	0.51 (± 0.02)
	Kurtosis	0.49 (± 0.07)
First-Order statistic	MeanAbsoluteDeviation	0.6 (± 0.08)
	Skewness	0.44 (± 0.1)
GLCM	ClusterProminence	0.41(± 0.4)
	ClusterShade	0.42 (± 0.44)

	ClusterTendency	0.54 (±0.25)
	DifferenceVariance	0.67(±0.13)
	SumAverage	0.61 (±0.001)
	SumSquares	0.57 (±0.2)
GLSM	LargeAreaLowGrayLevelEmphasis	0.39(±0.1)
	LowGrayLevelZoneEmphasis	0.4(±0.07)
	SmallAreaLowGrayLevelEmphasis	0.47(±0.07)
	GrayLevelVariance	0.58(±0.2)
GLRLM	HighGrayLevelRunEmphasis	0.63(±0.03)
	LongRunHighGrayLevelEmphasis	0.59(±0.04)
	LongRunLowGrayLevelEmphasis	0.33(±0.1)
	LowGrayLevelRunEmphasis	0.47(±0.1)
	ShortRunLowGrayLevelEmphasis	0.51(±0.08)
NGTDM	Busyness	0.6(±0.2)
	Strength	0.55(±0.1)
	GrayLevelVariance	0.58(±0.2)
GLDM	HighGrayLevelEmphasis	0.62(±0.02)
	LargeDependenceLowGrayLevelEmphasis	0.33(±0.1)
	LowGrayLevelEmphasis	0.48(±0.1)
	log-sigma-4-0-mm-3D_RobustMeanAbsoluteDeviation	0.66(±0.02)
LOG	log-sigma-4-0-mm-3D_Variance	0.57(±0.2)
	log-sigma-3-5-mm-3D_Kurtosis	0.69(±0.07)
	log-sigma-3-5-mm-3D_RobustMeanAbsoluteDeviation	0.64(±0.02)
	log-sigma-3-5-mm-3D_Variance	0.57(±0.2)
	log-sigma-2-5-mm-3D_Variance	0.61(±0.14)
	log-sigma-2-0-mm-3D_90Percentile	0.48(±0.1)
	log-sigma-1-5-mm-3D_90Percentile	0.34(±0.15)
	log-sigma-1-0-mm-3D_90Percentile	0.3(±0.2)
	wavelet-LHH_Mean	0.53(±0.1)
	wavelet-HLH_Mean	0.5(±0.2)
Wavelet	wavelet-HLH_Skewness	0.59(±0.03)
	wavelet-HHL_Mean	0.52(±0.2)
	wavelet-HHL_Median	0.52(±0.2)
	wavelet-HHH_Mean	0.24(±0.2)
	wavelet-HHH_Median	0.23(±0.2)

Table S10. More detailed information about the Radiomic features used in this study.

Radiomic Features Information		
Quantization method	Fixed-bin width	
Bin-width size	25	
Median bin counts (range) for original image	90 (40–140)	
Median bin counts (range) for all the filters	85 (30–120)	
Laplcn of Gaussian (LoG)	LoG with 10 sigma levels (0-5mm, strides of 0.5mm)	
	For each level, 18 features from first-order statistics were extracted	
Wavelet	Three channels	
	For each level, 18 features from first-order statistics were extracted	

Table S11. The hazard ratio for each covariate in the maximal cox proportional hazard model.

		Hazard ratio			
	Covariates added to model	BY	LS	MH	SK
		<i>p</i> -value	<i>p</i> -value	<i>p</i> -value	<i>p</i> -value
LUNG3 NSCLC-Radiomics- Genomics Harvard Dataset	PC1	1.0 (0.98-1.0) 0.8	1.00 (0.97-1.0) 0.8	0.99 (0.96-1.0) 0.3	1.00 (0.97-1.0) 0.9
	PC2	1.02(0.99-1.0) 0.1	1.03 (1.00-1.1) 0.03	1.02 (0.99-1.1) 0.1	1.04 (1.01-1.1) 0.007
	PC3	1.0 (0.95-1.0) 0.8	1.04 (1.00-1.1) 0.03	1.02 (0.96-1.1) 0.5	0.96 (0.92-1.0) 0.08
	Histology	0.92 (0.61-1.4) 0.7	0.83 (0.55-1.3) 0.3	0.87 (0.58-1.3) 0.5	0.69 (0.43-1.1) 0.1
	Stage	1.32 (0.91-1.9) 0.1	1.38 (0.97-2.0) 0.07	1.23 (0.85-1.8) 0.2	1.34 (0.93-1.9) 0.1
	Gender	1.32 (0.67-2.6) 0.4	1.36 (0.70-2.7) 0.3	1.3 (0.66-2.6) 0.4	1.32 (0.66-2.6) 0.4
NSCLC-Radioge- nomics Stanford Dataset	PC1	1.0 (0.98-1.0) 0.8	1.00 (0.97-1.0) 0.8	1.02 (0.99-1.0) 0.1	1.0 (0.98-1.0) 0.7
	PC2	1.06 (1.02-1.1) 0.003	1.03 (1.00-1.1) 0.03	1.03 (1.01-1.1) 0.01	1.1 (1.02-1.1) 0.001
	PC3	0.95 (0.90-1.0) 0.1	0.99 (0.95-1.0) 0.7	0.97 (0.92-1.0) 0.3	1.1 (1.00-1.1) 0.06
	Histology	0.83 (0.35-1.9) 0.5	0.92 (0.46-1.9) 0.8	0.94 (0.47-1.9) 0.8	1.0 (0.49-2.0) 0.9
	Morphology	0.82 (0.35-1.9) 0.6	1.06 (0.47-2.4) 0.8	0.95 (0.41-2.2) 0.9	1.5 (0.69-3.1) 0.3
	Gender	3.46 (1.41-8.5) 0.007	3.59 (1.46-8.8) 0.005	3.27 (1.33-8.1) 0.01	3.4 (1.35-8.3) 0.009