Dynamic Features refer to:

	ENANA
A.countor	EMM model parameter A
alpha.countor	EMM model parameter alpha
beta.countor	EMM model parameter beta
iAUC1.countor	V.1 feature (white paper)
Slope_ini.countor	V.2 feature (white paper)
Tpeak.countor	V.3 feature (white paper)
Kpeak.countor	V.5 feature (white paper)
SER.countor	V.4 feature (white paper)
maxCr.contour	III.2 feature (white paper)
peakCr.contour	IIII.3 feature (white paper)
UptakeRate.contour	IIII.4 feature (white paper)
washoutRate.contour	IIII.5 feature (white paper)
maxVr.contour	IV.4 feature (white paper)
peakVr.contour	IV.5 feature (white paper)
Vr_increasingRate.contour	IV.6 feature (white paper)
Vr_decreasingRate.contour	IV.7 feature (white paper)
Vr_post_1.contour	IV.8 feature (white paper)
minFri	Min iV.3
maxFri	Max IV.3
meanFri	Mean of IV.3
varFri	Variance of IV.3
skewFri	Skew of IV.3
kurtFri	Kurtosis of IV.3
iMaxVarianceuptake	<u>IV.1</u>
iiMin_change_Variance_uptake	<u>IV.2</u>
iiiMax_Margin_Gradient	<u>I.1</u>
k_for_Max_MarginGrad	k post of I.1
ivVariance	<u>1.2</u>

Notes: features on blue are also extracted from the inside voxels of the lesions (e.g A.inside)

Morphological Features refer to:

circularity	<u>1.3</u>
irregularity	<u>1.4</u>
edge_sharp_mean	From Levman et al. [1][8]
edge_sharp_std	From Levman et al. [8]
max_RGH_mean	Mean of I.5
max_RGH_mean_k	K post of I.5
max_RGH_var	Variance I.5

Textural Features refer to:

"texture_contrast_zero" "texture_contrast_quarterRad" "texture_contrast_halfRad" "texture_contrast_threeQuaRad" "texture_homogeneity_zero" "texture_homogeneity_quarterRad" texture homogeneity halfRad" "texture_homogeneity_threeQuaRad" "texture dissimilarity zero" "texture_dissimilarity_quarterRad" "texture dissimilarity halfRad" "texture dissimilarity threeQuaRad" "texture_correlation_zero" "texture_correlation_quarterRad" "texture_correlation_halfRad" "texture_correlation_threeQuaRad" "texture_ASM_zero" "texture ASM quarterRad" "texture_ASM_halfRad" "texture ASM threeQuaRad" "texture_energy_zero" "texture energy quarterRad" texture_energy_halfRad" "texture_energy_threeQuaRad"

Refers to texture features at different orientations O=[0,pi/4,pi/2,3pi/4]

Based on literature definition of on GLCM texture Haralick et al.

Labels:

table(set\$BenignNMaligNAnt)

massB massM nonmassB nonmassM 69 104 31 39

Time-intensity 3 parameter Empirical Model Fit proposed in [2]

The following model can be used to describe the lesion contrast uptake and washout and to fit the data:

$$\Delta S(t) = A \cdot (1 - e^{-\alpha t}) \cdot e^{-\beta t},$$

Where **A** is the upper limit of the signal intensity, **alpha(min⁻¹)** is the rate of signal increase, **beta(min⁻¹)** is the rate of the **signal decrease during washout**. The goodness of fit parameter R² will be calculated for each lesion.

#	Feature name	Description	Requires	Intuition
V.1	Initial area under curve (AUC)	$AUC_{\tau} = A \cdot [(1 - e^{-\beta \tau})/\beta]$	$+(e^{-(\alpha+\beta)\tau}-1)/(\alpha+\beta)],$	
	[2]	where tao is the time over w	hich signal intensity is	
		taken.		
V.2	Initial slope of enhancement	$Slope_{\mathrm{ini}} \approx A\alpha$.		
	(Slopeini)	litial slope of the kinetic		
	[2]	curve can be calculated by		
		taking the derivative of		
		AUC		
V.3	Time to peak of enhancement (Tpeak) [2]	$T_{\mathrm{peak}} = \frac{1}{\alpha} \log \left(1 - \frac{1}{\alpha} \right)$	$+\frac{\alpha}{\beta}$.	
		The time at which the kinetic	curve reached neak	
		Taken from setting the deriva	•	
V.4	Signal enhancement ratio (SER) [2]	$SER = \frac{\Delta S_1}{\Delta S_L} = \frac{1 - 1}{1 - 1}$	$\frac{e^{-\alpha t_1}}{e^{-\alpha t_L}} \cdot e^{(t_L - t_1)\beta}$	
		The signal intensity change at	the first timepoint (delta	
		S1) relative to the last timepo	oint (delta SL):	
V.5	Enhancement curvature at peak (kpeak)	$\kappa_{peak} \approx -A\alpha\beta$.		
	[2]	The curvature at the peak		
	[-]	of enhancement was		
		calculated from the		
		definition of curvature		
		formula at time of Tpeak		

Enhancement-kinetics features

#	Feature name	Description	Requires	Intuition
III.1	Definitions: contrast enhancement & average enhancement [3]	$C(\mathbf{r},i) = \frac{S(\mathbf{r},i) - S(\mathbf{r},0)}{S(\mathbf{r},0)}, i = 0,1,,5.$ In special circumstances this relation approaches linearity. $\bar{C}(i) = \frac{1}{L} \sum_{r=1}^{L} C(\mathbf{r},i), i = 0,1,,5,$	S(r,i) as the voxel value	related to Gd- DTPA concentration in the extracellular space of breast tissue at voxel

III.2	maximum uptake	$F_{II,1} = \max_{i=0,1,,5} \bar{C}(i)$	contrast enhancement &
	(FII,1)	and the state of t	average enhancement
	[3]		
III.3	peak location (FII,2)	time frame index at which the maximum	contrast enhancement &
	[3]	enhancement occurs	average enhancement
			& maximum uptake
111.4	uptake rate (FII,3) [3]	$F_{II,3} = F_{II,1}/F_{II,2}$.	
III.5	washout rate (FII,4) [3]	$F_{II,4} = \begin{cases} \frac{F_{II,1} - \bar{C}(5)}{5 - F_{II,2}} & (F_{II,2} \neq 5) \\ 0 & (F_{II,2} = 5) \end{cases}$	
		$(F_{II,2}=5)$	

Enhancement-variance dynamic features

	Limaneement-variance	aynamic reatares		
#	Feature name	Description	Requires	Intuition
IV.1	Max inhomogeneity of contrast uptake [4]	$\max_{i=0,,M-1} \left\{ \frac{\text{variance}_{\mathbf{r}} F_l(\mathbf{r},i)}{\text{variance}_{\mathbf{r}} F_l(\mathbf{r},0)} \right\}$ runs from frame 0 (before injection of contrast) to M-1 (M is the total number of time frames)	set of voxel values in the lesion at time frame "i" is given by F _I (r,i), where vectors r point to the lesion, and index "i"	
IV.2	Min variance of uptake, [4]	$\min_{i=0,,M-2} \left\{ \frac{\text{variance}_{\mathbf{r}} F_l(\mathbf{r},i)}{\text{variance}_{\mathbf{r}} F_l(\mathbf{r},i+1)} \right\},$	Variance F _I (r,i) is the computation of the variance of the voxel values at all r in the lesion	
IV.3	spatial variance of enhancement [3]	$V(i) = \frac{1}{L-1} \sum_{r=1}^{L} \left[C(\mathbf{r}, i) - \bar{C}(i) \right]^{2}, i = 0$		describes the time course of the spatial variance of
		Five features are derived from the enhanced dynamics. The	ment-variance	the enhancement within the lesion
IV.4	maximum variation of	Maximum spatial variance of		
	enhancement (FIII,1)	enhancement $F_{III,1} = \max_{i=0,1,,5} V(i)$		
IV.5	peak location (FIII,2) of enhancement	time frame index at which the maximum variance occurs		
IV.6	enhancement-variance increasing rate (FIII,3)	$F_{III,3} = F_{III,1}/F_{III,2}$		how fast the enhancement- variation within the lesion reaches maximum

IV.7	enhancement-variance decreasing rate (FIII,4)	$F_{III,4} = \begin{cases} \frac{F_{III,1} - V(5)}{5 - F_{III,2}} \\ 0 \end{cases}$	$(F_{III,2} \neq 5)$ $(F_{III,2} = 5)$	how fast the enhancement- variance decreases from the maximum
IV.8	Enhancement-variance at the first postcontrast frame (FIII,5),	V(1)		reveals the uptake inhomogeneity at the early phase of uptake.

Feature name	Description	Requires	Intuition
3D Sharpness of lesion margin [4]	$\max_{i=0,,M-1} \left\{ \frac{\mathrm{mean_r} \ \nabla [F_m(\mathbf{r},i) - F_m(\mathbf{r},0)]\ }{\mathrm{mean_r} F_m(\mathbf{r},i)} \right\}$ Fm is limited to a shell—three voxels thick—centered on the surface of the lesion.	$\nabla [F_m(\mathbf{r},i) - F_m(\mathbf{r},0)]$ is the set of voxel-value gradients at the margin of the suspect lesion in the difference images of time frame "i" and precontrast frame "0." sharpness of the uptake of contrast is computed at the lesion margin.	
3D variance of margin gradient [4]	$\frac{\text{variance}_{\mathbf{r}} \ \nabla [F_m(\mathbf{r},i) - F_m(\mathbf{r},0)]\ }{[\text{mean}_{\mathbf{r}} F_m(\mathbf{r},i)]^2}$ Computation of the spatial gradient is accomplished in 3D by convolution with the components of a 3x3x3 Sobel filter in three orthogonal directions.	only computed from the subtraction frames of "i" and "0" where the margin Sharpness of lesion margin is maximum.	malignant lesions take up contrast agent in a less homogeneous pattern than benign masses

Morphologic features

#	Feature name	Description	Requires	Intuition
1.3	3 3D Circularity [4]	volume of lesion within sphere of effective diameter volume of lesion	Volume of lesion and effective diameter	"Circularity" quantifies how well the lesion conforms to a spherical shape
1.4	S3D Irregularity [4]	$1 - \frac{\pi \cdot \text{effective diameter}^2}{\text{surface of lesion}}$ Effective diameter = $2 \cdot \sqrt[3]{\frac{3 \cdot \text{volume of lesion}}{4\pi}}$	Surface of lesion and effective diameter.	malignant lesions have less sharp boundaries and are more

		circularity and irregularity in 3D are computed	I	irregularly
		in world coordinates— rather than in voxel	shaped.	
		coordinates—to account for anisotropic		"irregularity"
		voxels.		indicates the
				roughness of
				the surface
1.5	radial	Dot product between voxel-value gradients		Radial
	gradient	and lines intersecting a single point near the		gradient
	histogram	center of the suspect lesion (lines in radial		analysis
	(RGH)	directions)		indicates how
	[4]	histogram—quantifying the frequency of		well
		occurrence of the dot products		structures in a
		occurrence of the dot products		ROI extend in
				a radial
				pattern
				originating
				from the
				center of the
				ROI.
-		may Systiance H(n)	Flat in a set of consolication	
1.6	variance of RGH	$\max_{i=0,,M-1} \{ \text{variance}_p H(p) \},\$	Fb is set of voxel-value	The variance of RGH values
			gradients in a rectangular	
	values. [4]	$p = \frac{\left \nabla [F_b(\mathbf{r}, i) - F_b(\mathbf{r}, 0)] \cdot (\mathbf{r} - \mathbf{r}_c)\right }{\left\ \nabla [F_b(\mathbf{r}, i) - F_b(\mathbf{r}, 0)]\right\ \cdot \left\ \mathbf{r} - \mathbf{r}_c\right\ }$	box encompassing the	is used to
		$p = \frac{\ \nabla [F_b(\mathbf{r},i) - F_b(\mathbf{r},0)]\ \cdot \ \mathbf{r} - \mathbf{r}_c\ }{\ \nabla [F_b(\mathbf{r},i) - F_b(\mathbf{r},0)]\ \cdot \ \mathbf{r} - \mathbf{r}_c\ }$	suspect lesion with an	quantify the
		- -	additional margin of 3	flatness of the
			voxels	RGH.
			normalized RGH	
I.7	max			
	standard			
	deviation			
	of RGH [3]			
1.8	Margin	$RSI_{in}(r_i, t) - RSI_{out}(n(r_i), t)$		
	Sharpness	1		
	[5]	a (i)		
		a function of position (ri) and time (t); ri is a vo	•	
		three-dimensional coordinates (the inner ring)		
		three-dimensional neighborhood.; tf is the fine	al time point of the	
		examination		
1		d is the normalization term		

Texture Features – based on GLCM texture Haralick et al. [6] [7]

For an image of **G** gray levels, the **G**×**G** gray level co-occurrence matrix Pd for a displacement vector **d** is defined as follows: The **entry** (**i**, **j**)**of** P**d** is the number of **occurrence of voxel-pair of gray levels i and j** whose spatial locations are a vector **d apart**. When normalized by the total counts, the **entry** (**i**, **j**)**of** P**d**, denoted **as p**(**i**, **j**), represents the (empirical) **probability of occurrence of voxel pair of gray levels i and j**

whose spatial locations are a vector d apart. For a particular **voxel (gray), it has 26 neighbors** of 1 voxel distance **d** in 13 independent directions.

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