

## 6 Appendix

**Table 3.** Clinical characteristics of patients in the dataset.

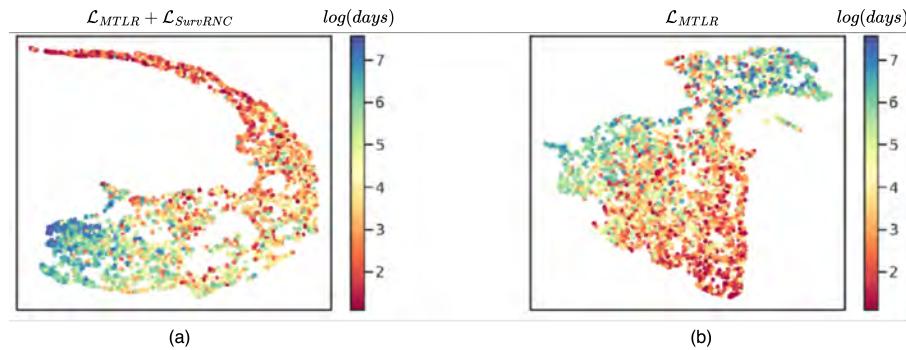
Characteristics	Dataset (n = 488)
<b>RFS</b>	
Uncensored	96 (19.7)
Censored	392 (80.3)
<b>Age (year)</b>	
<b>Weight (kg)</b>	
<b>Gender</b>	
Male	402 (82.4)
Female	86 (17.6)
<b>Alcohol consumption</b>	
Yes	95 (19.5)
No	59 (12.1)
Unknown	334 (68.4)
<b>Tobacco consumption</b>	
Yes	85 (17.4)
No	105 (21.5)
Unknown	298 (61.1)
<b>HPV status</b>	
Positive	274 (56.1)
Negative	43 (8.8)
Unknown	171 (35.1)
<b>Performance status</b>	
0	86 (17.7)
1	114 (23.3)
2	11 (2.3)
3	3 (0.6)
4	1 (0.2)
Unknown	273 (55.9)
<b>Surgery</b>	
Yes	50 (10.3)
No	248 (50.8)
Unknown	190 (38.9)
<b>Chemotherapy</b>	
Yes	422 (86.5)
No	66 (13.5)

**Table 4.** Preprocessing and augmentation details.

Augmentations	Axis	Probability	Size
Orientation	PLS	-	-
CT/PET Concatenation	1	-	-
Normalization	-	-	-
Random crop	-	0.5	96 x 96 x 96
Random flip	x, y, z	0.1	-
Rotate by 90 (up to 3x)	x, y	0.1	-

**Table 5.** Ablation to study the effect of the parameter  $\lambda$  values.

Models	$\lambda$			
	0.3	0.5	0.7	1.0
DeepMTLR [8] $\mathcal{L}_{MTLR} + \mathcal{L}_{SurvRNC}$	0.6870	0.7009	0.6690	0.6742
DeepHit [14] $\mathcal{L}_{Hit} + \mathcal{L}_{SurvRNC}$	0.6802	0.7233	0.6950	0.6808

**Fig. 3. UMAP Visualization:**  $\mathcal{L}_{SurvRNC}$  with the native loss function of a DeepMTLR survival model on the SUPPORT dataset [13] shows a better continuous latent feature representation (a) as compared to using only the native loss function (b).