

Tissue Phenomics for prognostic biomarker discovery in low- and intermediate-risk prostate cancer

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-- Supplemental Material --

Supplemental Figures

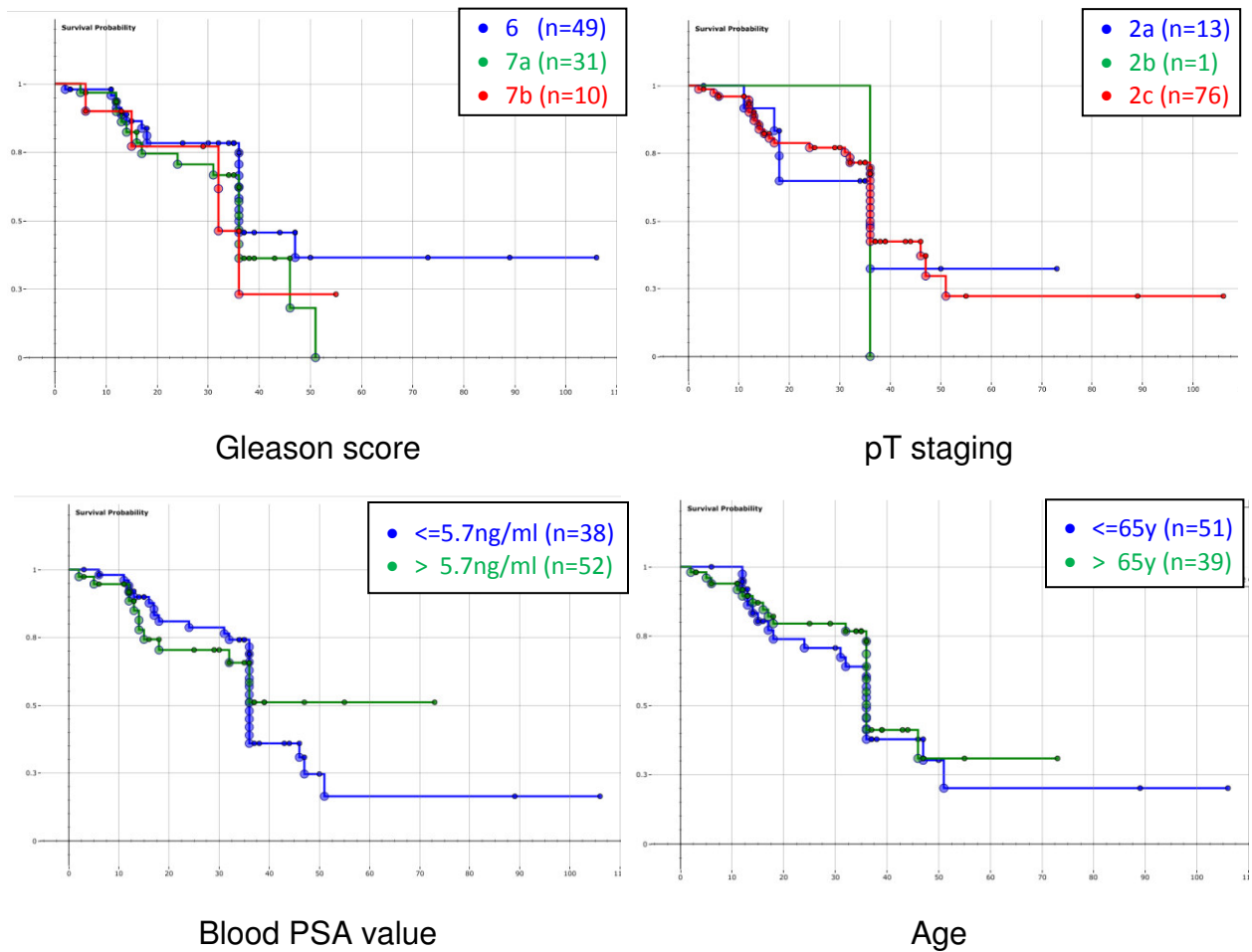


Figure S1: Stratification based on standard clinical data with number of samples per group n in braces. For the blood PSA value and age the cut-off has been optimized to maximize the log-rank test p-value.

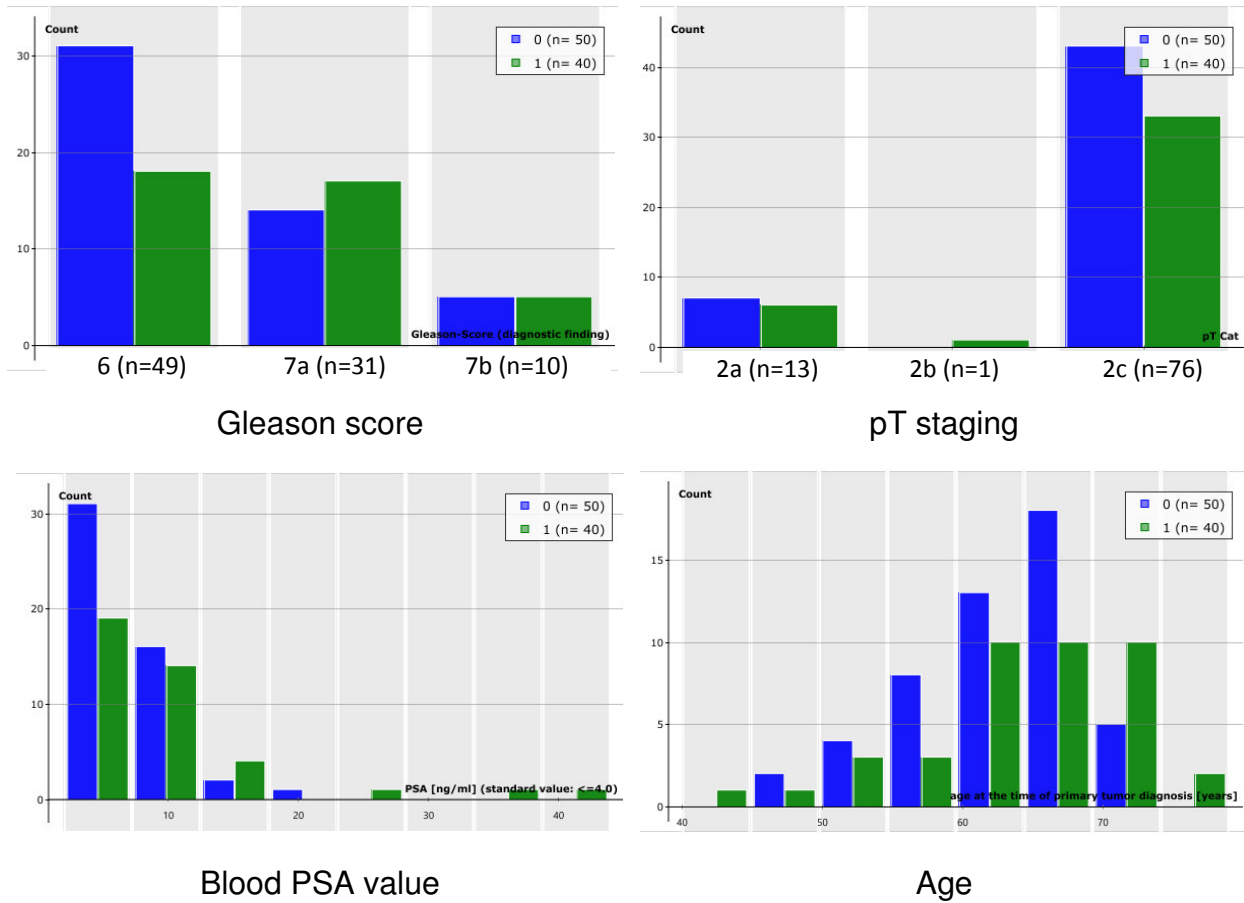
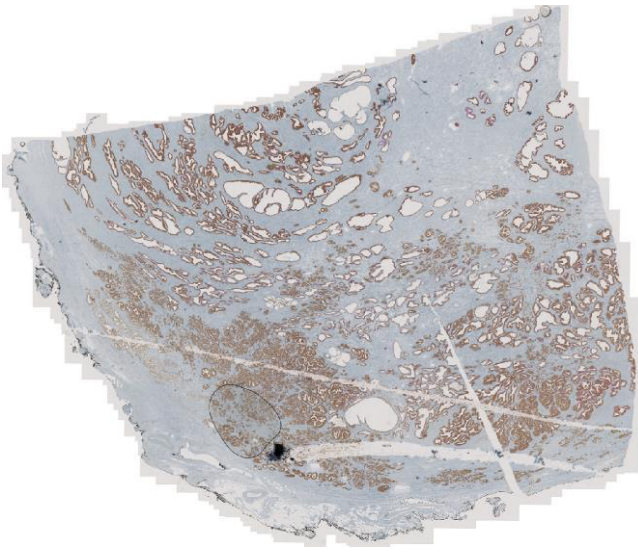
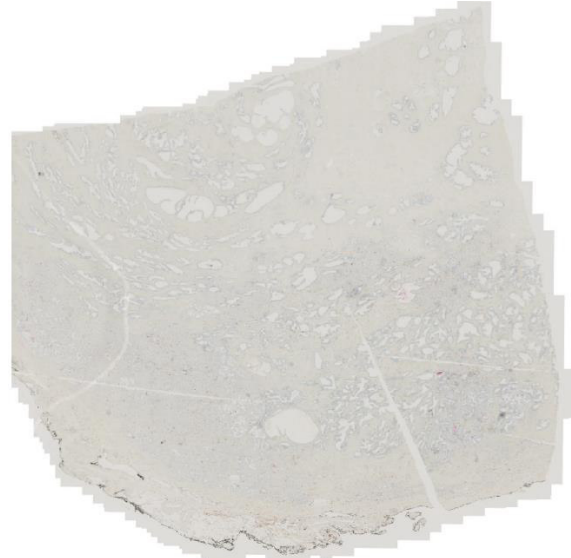


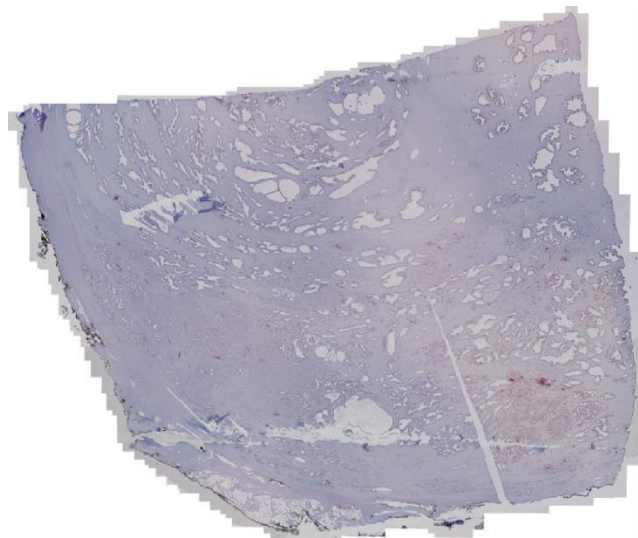
Figure S2: Histograms of standard clinical data over all patients where 0 (blue) refers to the patients without tumor progression and 1 (green) to the patients with tumor progression.



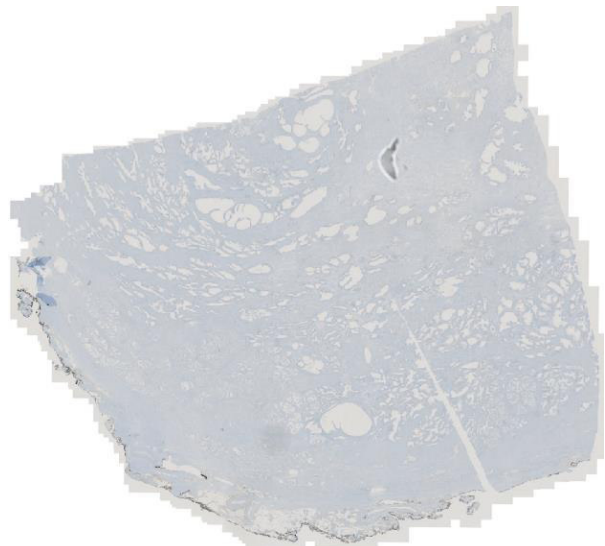
CK18/p63



CD68/CD163

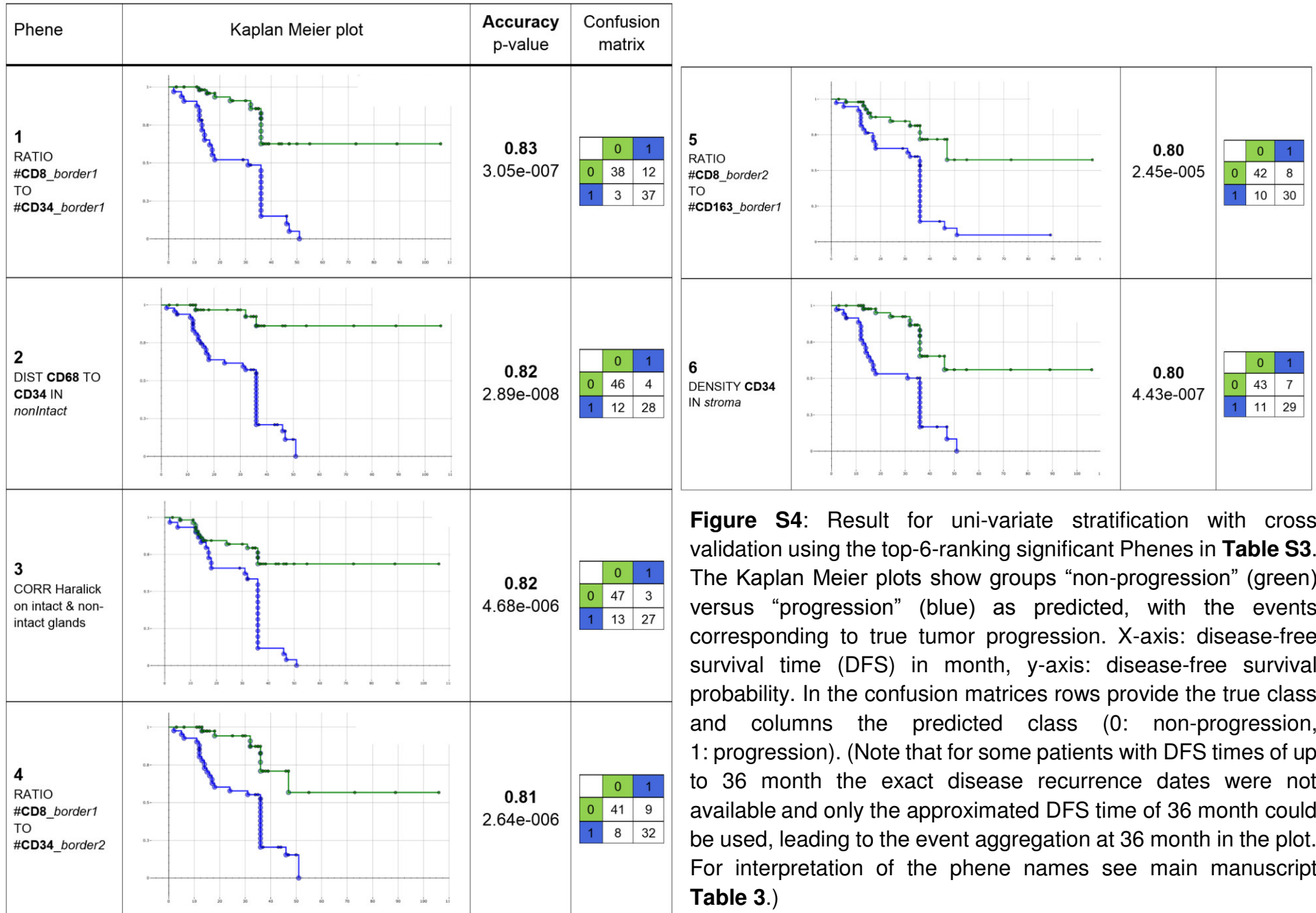


CD3/CD8



CD34

Figure S3: Examples of whole-slide images of consecutive FFPE tissue sections from resected *prostate cancer* tissue.



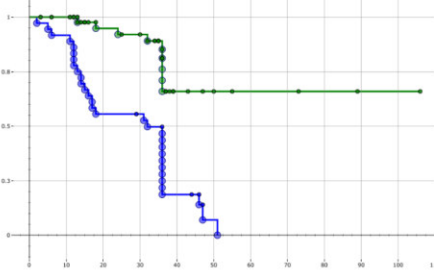
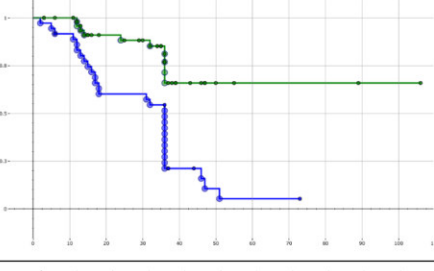
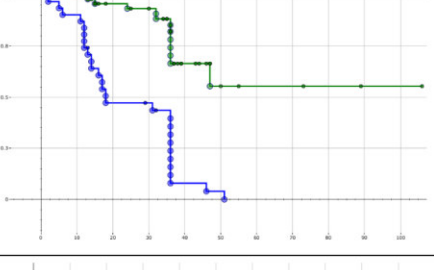
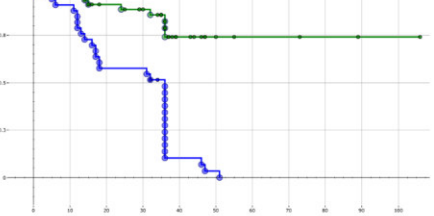
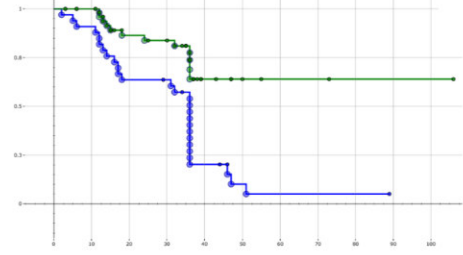
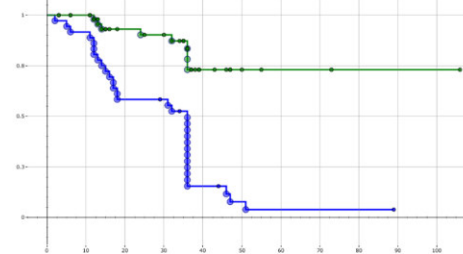
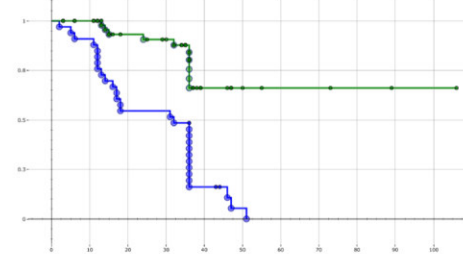
Method	Kaplan Meier plot	Accuracy p-value	Confusion matrix									
hclust (ward)		0.84 1.25e-007	<table><tr><td></td><td>0</td><td>1</td></tr><tr><td>0</td><td>45</td><td>5</td></tr><tr><td>1</td><td>9</td><td>31</td></tr></table>		0	1	0	45	5	1	9	31
	0	1										
0	45	5										
1	9	31										
Bayes		0.82 7.95e-006	<table><tr><td></td><td>0</td><td>1</td></tr><tr><td>0</td><td>44</td><td>6</td></tr><tr><td>1</td><td>10</td><td>30</td></tr></table>		0	1	0	44	6	1	10	30
	0	1										
0	44	6										
1	10	30										
CART		0.83 2.69e-010	<table><tr><td></td><td>0</td><td>1</td></tr><tr><td>0</td><td>47</td><td>3</td></tr><tr><td>1</td><td>12</td><td>28</td></tr></table>		0	1	0	47	3	1	12	28
	0	1										
0	47	3										
1	12	28										
kNN (k=5)		0.88 3.32e-009	<table><tr><td></td><td>0</td><td>1</td></tr><tr><td>0</td><td>48</td><td>2</td></tr><tr><td>1</td><td>9</td><td>31</td></tr></table>		0	1	0	48	2	1	9	31
	0	1										
0	48	2										
1	9	31										
Linear Predictor		0.81 2.46e-005	<table><tr><td></td><td>0</td><td>1</td></tr><tr><td>0</td><td>45</td><td>5</td></tr><tr><td>1</td><td>12</td><td>28</td></tr></table>		0	1	0	45	5	1	12	28
	0	1										
0	45	5										
1	12	28										
SVM (lin. kernel)		0.87 5.47e-008	<table><tr><td></td><td>0</td><td>1</td></tr><tr><td>0</td><td>46</td><td>4</td></tr><tr><td>1</td><td>8</td><td>32</td></tr></table>		0	1	0	46	4	1	8	32
	0	1										
0	46	4										
1	8	32										
SVM (rbf kernel)		0.86 3.55e-008	<table><tr><td></td><td>0</td><td>1</td></tr><tr><td>0</td><td>47</td><td>3</td></tr><tr><td>1</td><td>10</td><td>30</td></tr></table>		0	1	0	47	3	1	10	30
	0	1										
0	47	3										
1	10	30										

Figure S5: Performance evaluation of multi-variate stratification using different types of classifiers with the subset of significant Phenes selected using the FWER method of Holm-Bonferroni (**Table S3**). The Kaplan Meier plots show groups “non-progression” (green) versus “progression” (blue) as predicted, with the events corresponding to true tumor progression. X-axis: disease-free survival time (DFS) in month, y-axis: disease-free survival probability. In the confusion matrices rows provide the true class and columns provide the predicted class (0 = non-progression, 1 = progression).

Figure S5: Performance evaluation of multi-variate stratification using different types of classifiers with the subset of significant Phenes selected using the FWER method of Holm-Bonferroni (**Table S3**). The Kaplan Meier plots show groups “non-progression” (green) versus “progression” (blue) as predicted, with the events corresponding to true tumor progression. X-axis: disease-free survival time (DFS) in month, y-axis: disease-free survival probability. In the confusion matrices rows provide the true class and columns the predicted class (0: non-progression, 1: progression). (Exact disease recurrence dates not available for some patients and therefore the approximated DFS time of 36 month is used.)

Supplemental Tables

	n	Uni-variate			Multi-variate		
		HR	95% CI	p-value	HR	95% CI	p-value
Gleason Score							
6	49						
7a	31	1.534	0.788 - 2.987	0.208	1.488	0.714 - 3.103	0.289
7b	10	1.610	0.596 - 4.346	0.348	1.815	0.646 - 5.103	0.258
pT							
2a	13						
2b	1	1.582	0.189 - 13.239	0.672	2.648	0.199 - 35.212	0.461
2c	76	0.865	0.361 - 2.077	0.746	0.660	0.256 - 1.705	0.391
Age	90	1.020	0.972 - 1.070	0.421	1.024	0.962 - 1.090	0.450
PSA blood value	90	1.024	0.984 - 1.066	0.251	1.010	0.964 - 1.058	0.674

Table S1: Result of Cox Regression on clinical features.

IHC	Antibody			Detection systems ¹	Pretreatment ¹
	Provider	Dilution	Incubation		
CD68/ CD163	Dako/ Ventana	1:23/ ready to use	40 min/ 16 min	ultraView AP Red/ optiView DAB	32 min, CC1
CD3/ CD8	Zytomed/ Cell Marque	1:150/ 1:50	60 min/ 24 min	ultraView AP Red/ ultraView DAB	60 min, CC1
CD34	Cell Marque	1:200	20 min	ultraView DAB	30 min, CC1
CD18/ p63	Progen / Zytomed	1:200/ 1:200	12 min/ 32 min	ultraView DAB/ ultraView AP Red	60 min, CC1

Table S2: Basic data of IHC protocols

¹ (Ventana Medical Systems, Tucson, Arizona)

	<u>Model</u> : tumor non-progression if									
	Phene	OP	THR	ACC	P-VAL	NUM 5	TP	TN	FP	FN
1	RATIO #CD8_border1 TO #CD34_border1	>=	0.10	0.83	3.05e-007	81	37	38	12	3
2	DIST CD68 TO CD34 IN nonIntact	>=	75.70	0.82	2.89e-008	13	28	46	4	12
3	CORR Haralick on intact & non-intact glands	<	0.09	0.82	4.68e-006	0	27	47	3	13
4	RATIO #CD8_border1 TO #CD34_border2	>=	0.03	0.81	2.64e-006	0	32	41	9	8
5	RATIO #CD8_border2 TO #CD163_border1	>=	0.65	0.80	2.45e-005	90	30	42	8	10
6	DENSITY CD34 IN stroma	<	0.72	0.80	4.43e-007	0	29	43	7	11
7	DENSITY CD34 IN nonIntact	<	0.85	0.80	2.20e-006	0	28	44	6	12
8	RATIO #CD8_stroma TO #CD34_stroma	>=	0.11	0.79	5.19e-006	0	29	42	8	11
9	RATIO #CD68_border2 TO #CD34_innerBorder	>=	0.07	0.78	1.16e-006	0	28	42	8	12
10	RATIO #CD3_stroma TO #CD34_border1	>=	0.45	0.77	0.00038	0	27	42	8	13
11	RATIO #CD163_stroma TO #CD34_innerBorder	>=	2.40	0.76	3.59e-006	0	31	37	13	9
12	RATIO #CD68_border2 TO #CD34_border1	>=	0.04	0.76	1.84e-005	0	25	43	7	15
13	RATIO #CD8_nonIntact TO #CD163_nonIntact	>=	0.40	0.74	0.00021	2	28	39	11	12
14	DIST CD3 TO CD8 IN innerBorder	<	98.69	0.74	0.00047	0	33	34	16	7
15	DIST CD3 TO CD8 IN border2	<	99.45	0.73	9.71e-005	66	31	35	15	9
16	RATIO #CD3_nonIntact TO #CD34_nonIntact	>=	0.18	0.73	0.00034	0	37	29	21	3
17	DIST CD163 TO CD8 IN intact	<	150.46	0.72	3.15e-005	0	25	40	10	15
18	RATIO #CD8_border2 TO #CD3_nonIntact	>=	1.57	0.72	0.00012	0	20	45	5	20
19	RATIO #CD8_ws TO #CD3_ws	>=	0.73	0.72	7.80e-005	0	22	43	7	18
20	RATIO #CD163_ws TO #CD34_stroma	>=	0.93	0.68	0.00022	0	36	25	25	4

Table S3: Ranked list of selected significant Phenes resulting from feature reduction using filtering for accuracy and correlation, followed by FWER multiple-testing correction (Holm 1979). Column **Phene** provides a short definition (for interpretation of the phene names see naming conventions below), which together with the operator **OP** and the threshold **THR** defines the model for classifying patients as tumor non-progression (distances are given in μm). **ACC** provides the cross-validated accuracy, **P-VAL** is the log-rank test p-value of the aggregated prediction, **NUM 5** represents the number of top-5 occurrences over all cross validation runs, and **TP**, **TN**, **FP**, and **FN** gives the true positive, true negative, false positive and false negative predictions, respectively. (For interpretation of the phene names see main manuscript **Table 3**.)

Selected phenes	Uni-variate			Multi-variate		
	HR	95% CI	p-value	HR	95% CI	p-value
RATIO #CD8_ <i>border1</i> TO #CD34_ <i>border1</i>	8.1e-06	1.3e-08 - 0.0051	0.00036	8.9e-04	1.2e-06 - 0.65	0.037
DIST CD68 TO CD34 IN <i>nonIntact</i>	0.99	0.99 - 1	6.7e-05	0.998	0.996 - 1.0	0.150
CORR Haralick on intact & non-intact glands	770	61 - 9700	2.7e-07	80.1	4.2 - 1531.2	0.004
RATIO #CD8_ <i>nonIntact</i> TO #CD163_ <i>nonIntact</i>	0.33	0.13 - 0.82	0.017			
DIST CD3 TO CD8 IN <i>border2</i>	1	1-1	5.5e-06			
RATIO #CD68_ <i>border2</i> TO #CD34_ <i>border1</i>	0.0025	2.1e-05 - 0.3	0.014			
DIST CD3 TO CD8 IN <i>innerBorder</i>	1	1 - 1	7.5e-06			
RATIO #CD8_ <i>ws</i> TO #CD3_ <i>ws</i>	0.85	0.61 - 1.2	0.340			

A

	n	Multi-variate		
		HR	95% CI	p-value
RATIO #CD8_ <i>border1</i> TO #CD34_ <i>border1</i>	90	0.001	7.2e-07 - 1.239	0.057
DIST CD68 TO CD34 IN <i>nonIntact</i>	90	0.997	0.993 - 1.000	0.034
CORR Haralick on intact & non-intact glands	90	149.9	5.192 - 4330.060	0.004
Gleason Score				
6	49			
7a	31	1.110	0.526 - 2.320	0.792
7b	10	2.452	0.832 - 7.229	0.104
pT				
2a	13			
2b	1	0.417	0.031 - 5.558	0.508
2c	76	0.533	0.202 - 1.408	0.204
Age	90	0.997	0.940 - 1.056	0.905
PSA blood value	90	0.981	0.939 - 1.026	0.407

B

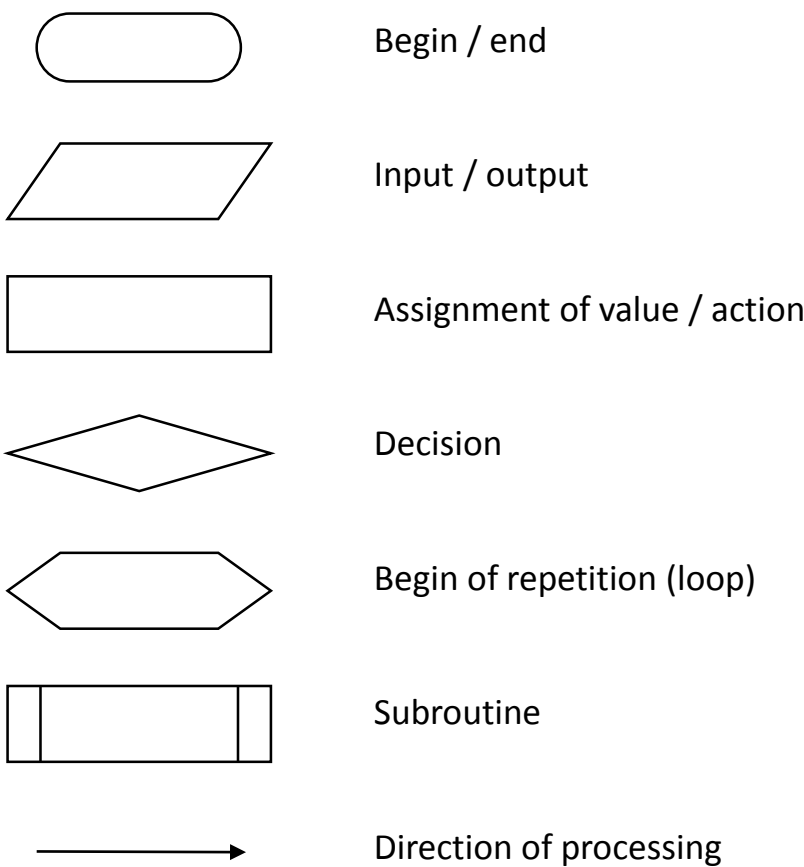
Table S4: Result of Cox regression for selected best phenes. (A) Uni-variate Cox regression (left) and multi-variate Cox regression (right, for the top 3 phenes only due to limited number of 40 events). (B) Multi-variate Cox regression of top 3 phenes together with the clinical features. (For interpretation of the phone names see **Table 3.**)

Algorithm Flow Charts

Table of Contents

- 1) Gland segmentation and classification
- 2) Cell nucleus segmentation
- 3) Image co-registration
- 4) Image feature extraction
- 5) Feature ranking and phene discovery

Flow Chart Symbols



References

[1] Brieu, N., Pauly, O., Zimmermann, J., Binnig, G., Schmidt, G. "Slide specific models for segmentation of differently stained digital histopathology whole slide images", in *SPIE Med. Imaging* 2016.

[2] Criminisi, A., Shotton, J., Bucciarelli, S., "Decision forests with long-range spatial context for organ localization in CT volumes," in *Medical Image Computing and Computer-Assisted Intervention (MICCAI)* 2009, pp. 69–80.

[3] Forssen, P.-E., "Maximally stable colour regions for recognition and matching," in *Computer Vision and Pattern Recognition (CVPR)* 2007, pp. 1–8.

[4] Sheeba, F., Thamburaj, R., Mammen, J. J., Nagar, A. K., "Splitting of overlapping cells in peripheral blood smear images by concavity analysis," in *Combinatorial Image Analysis*, pp. 238–249 (2014).

[5] Yigitsoy, M. & Schmidt, G. "Hierarchical patch-based co-registration of differently stained histopathology slides", in *SPIE Med. Imaging* 2017.

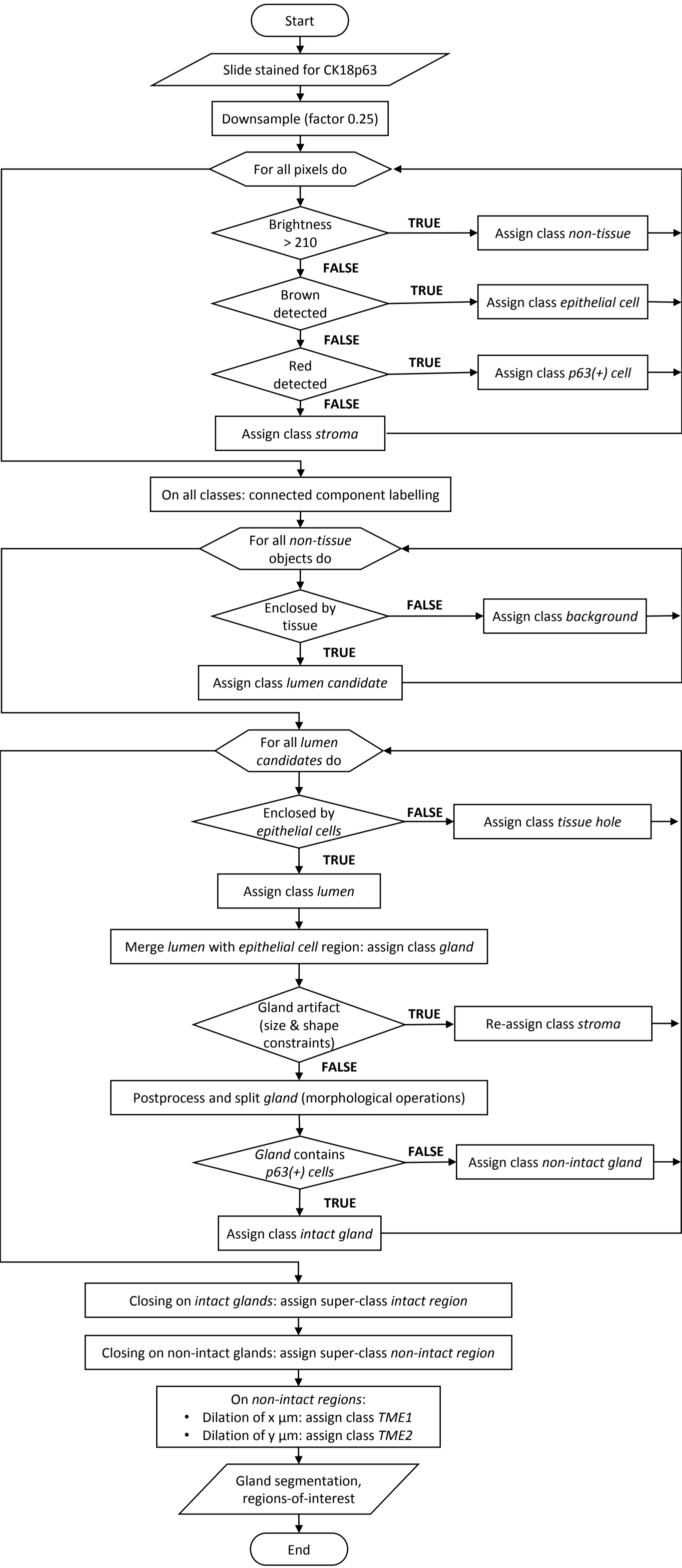
[6] Zhu, C., Byrd, R. H., Lu, P., Nocedal, J., "Algorithm 778: L-bfgs-b: Fortran subroutines for large-scale bound-constrained optimization," in *ACM Transactions on Mathematical Software (TOMS)*, 23(4), pp. 550-560 (1997).

[7] Harder, N. et al. "Co-occurrence features characterizing gland distribution patterns as new prognostic markers in prostate cancer whole-slide images", in *International Symposium on Biomedical Imaging (ISBI)* 2016, pp. 807–810.

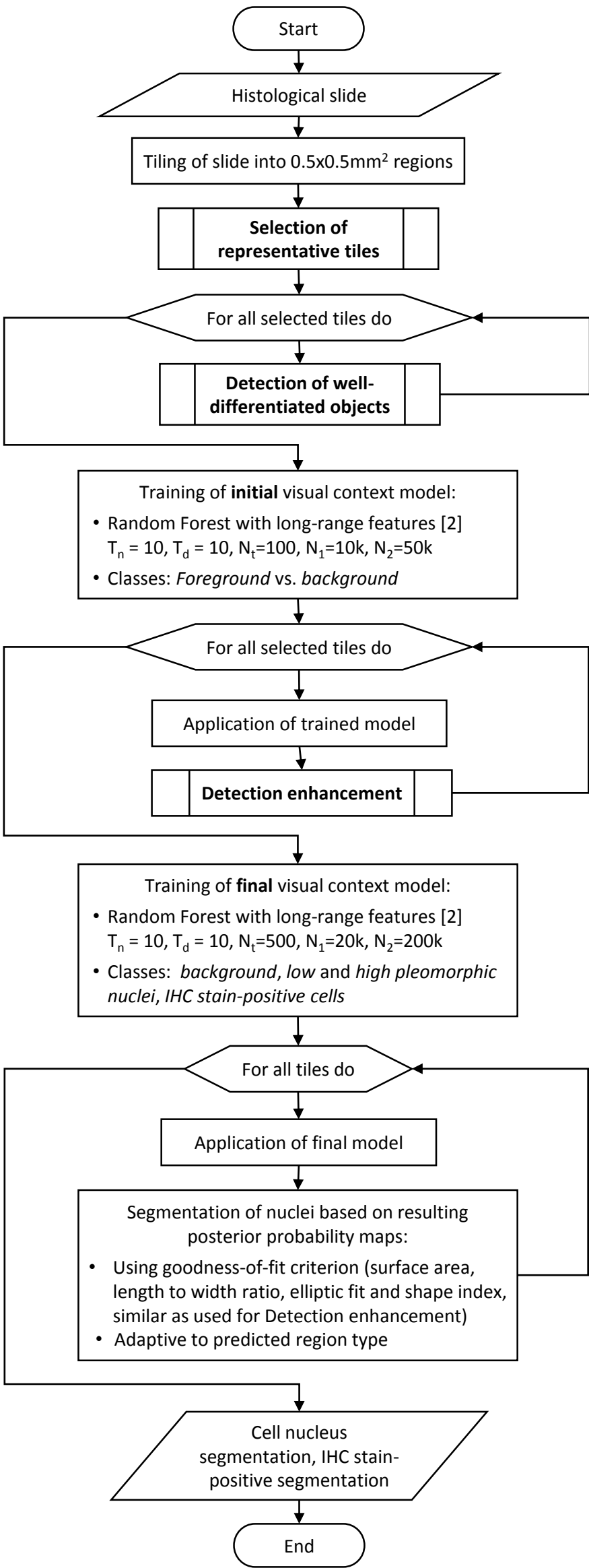
[8] Holm, S. "A Simple Sequentially Rejective Multiple Test Procedure", *Scand. J. Stat.* 6, 65–70 (1979).

[9] Breiman, L., Friedman, J., Stone, C. J. & Olshen, R. A. "Classification and Regression Trees", (Taylor & Francis, 1984).

1) Gland segmentation and classification



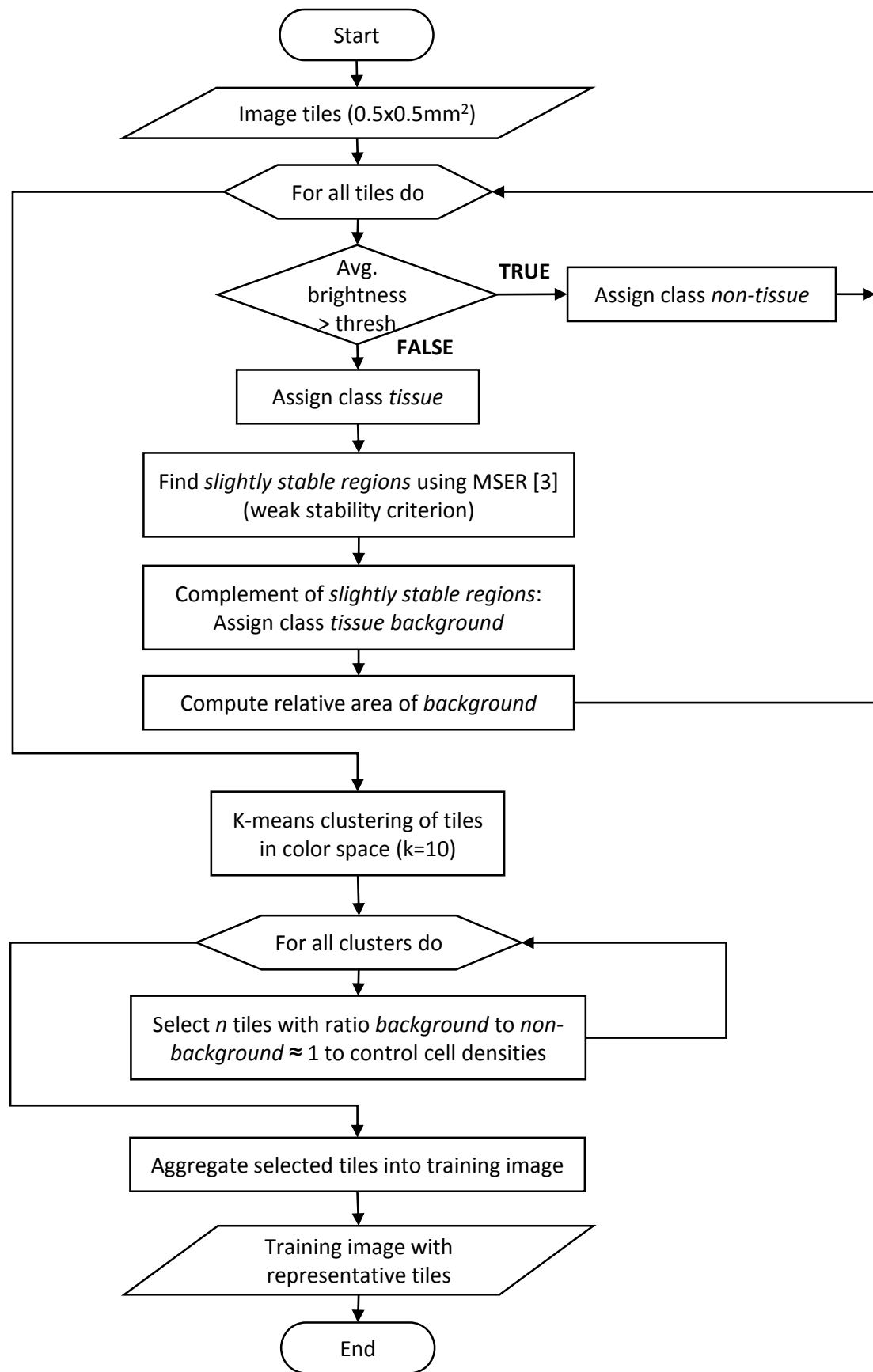
2) Cell nucleus segmentation [1]



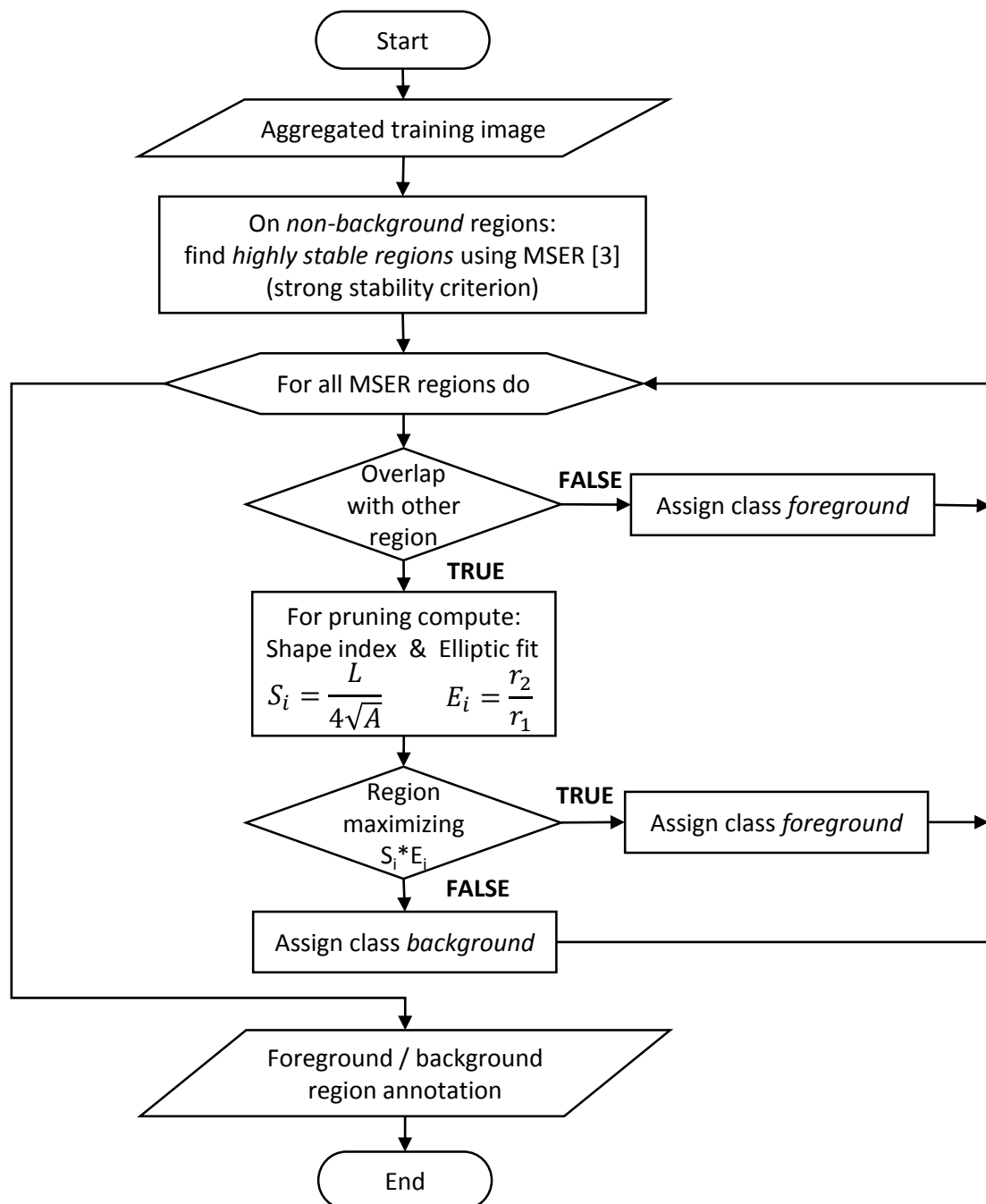
Random forest parameters

- T_n : number of trees
- T_d : tree depth
- N_t : number of tries
- N₁: number of random samples used to learn the structure of the random forest
- N₂: number of random samples used to refine the thresholds at each node

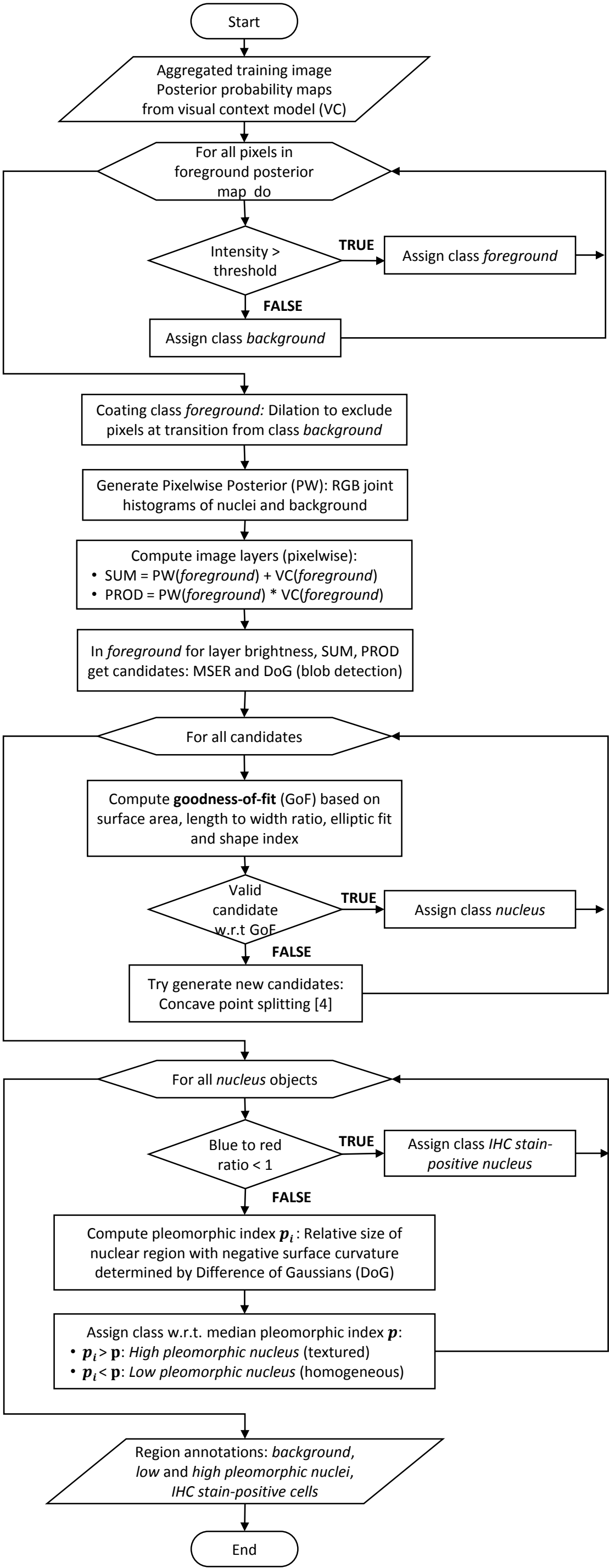
2.1) Cell nucleus segmentation: Selection of representative tiles



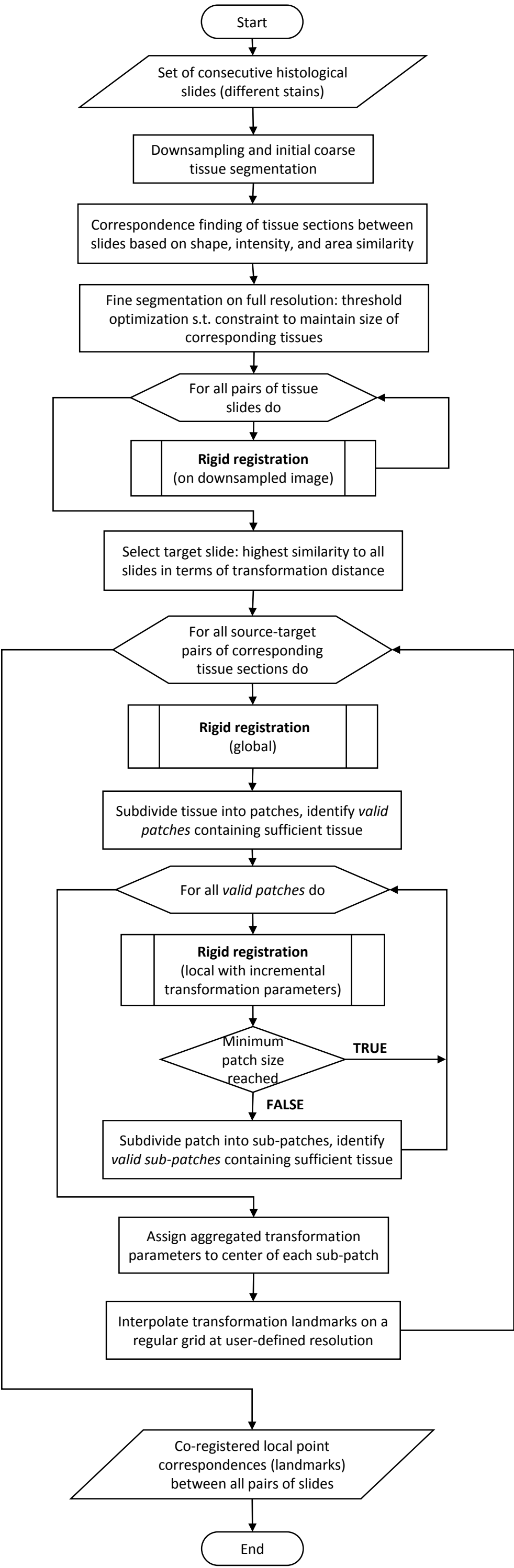
2.2) Cell nucleus segmentation: Detection of well-differentiated objects



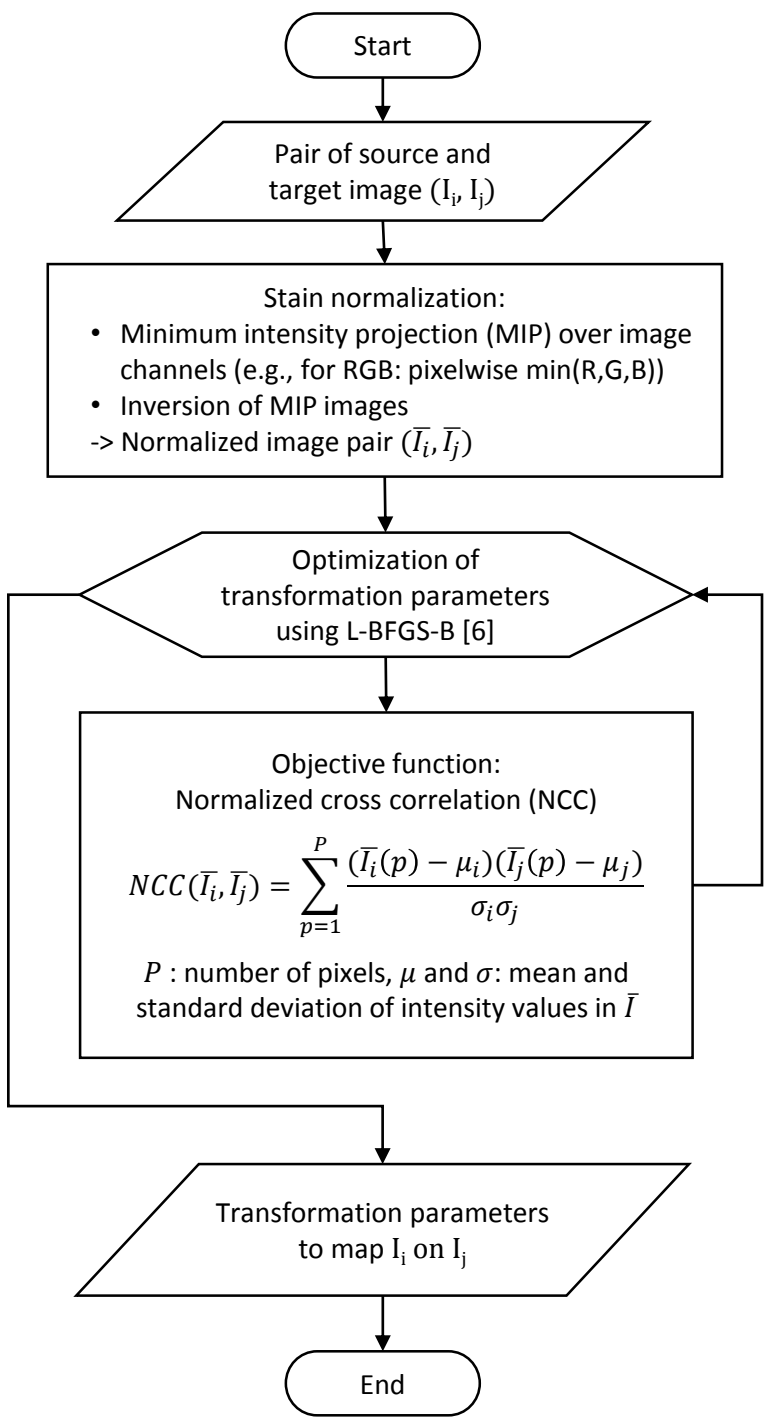
2.3) Cell nucleus segmentation: Detection enhancement



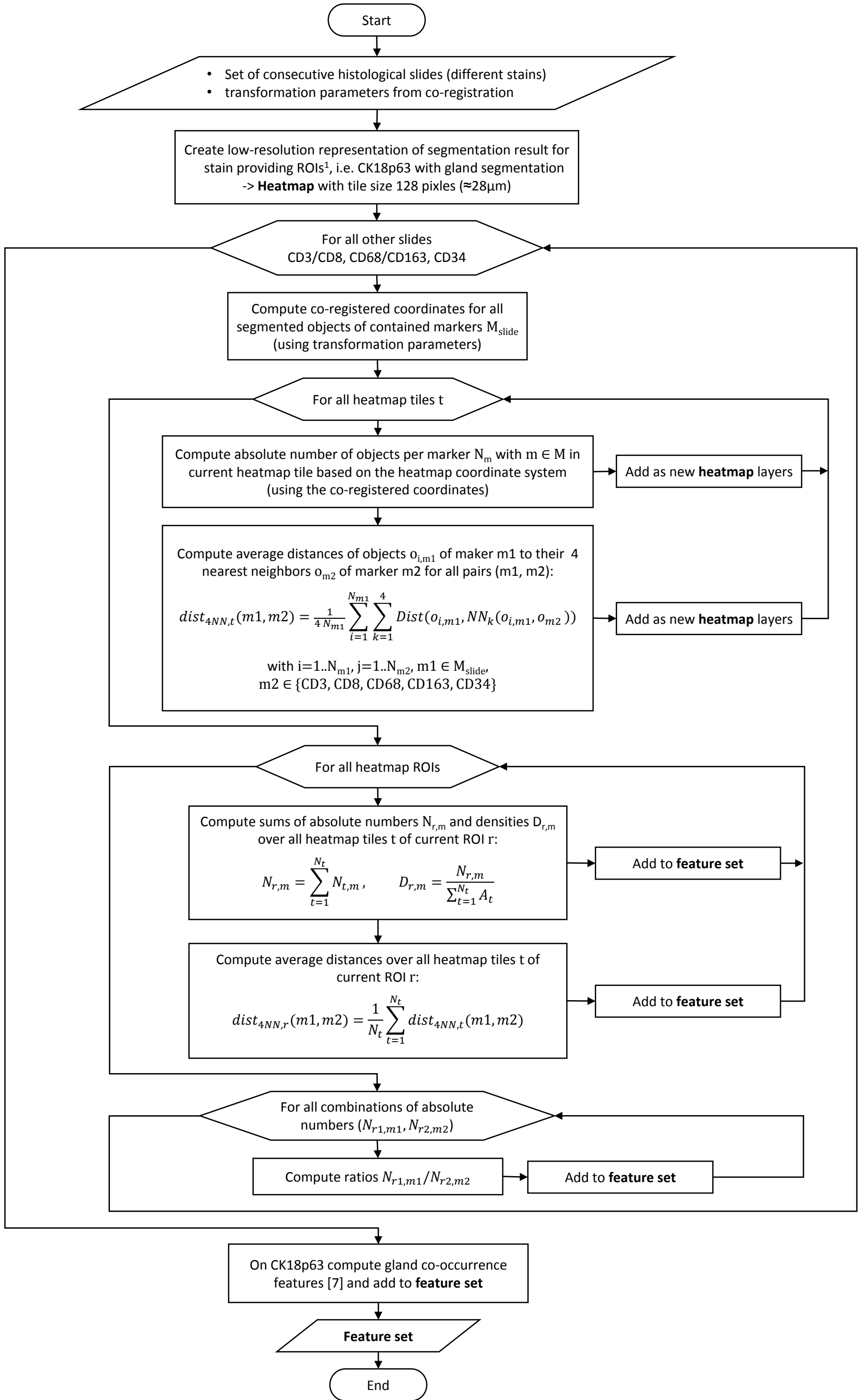
3) Image co-registration [5]



3.1) Image co-registration: Rigid registration

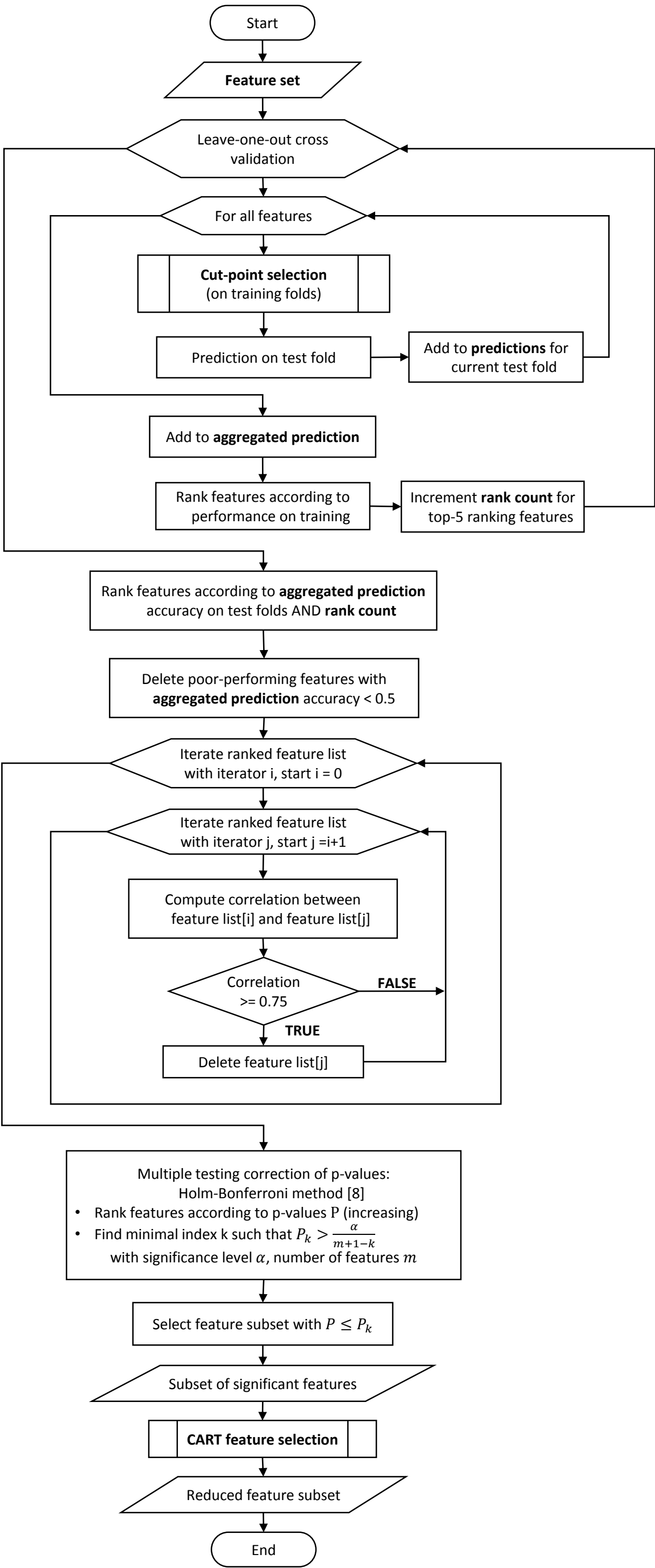


4) Image feature extraction

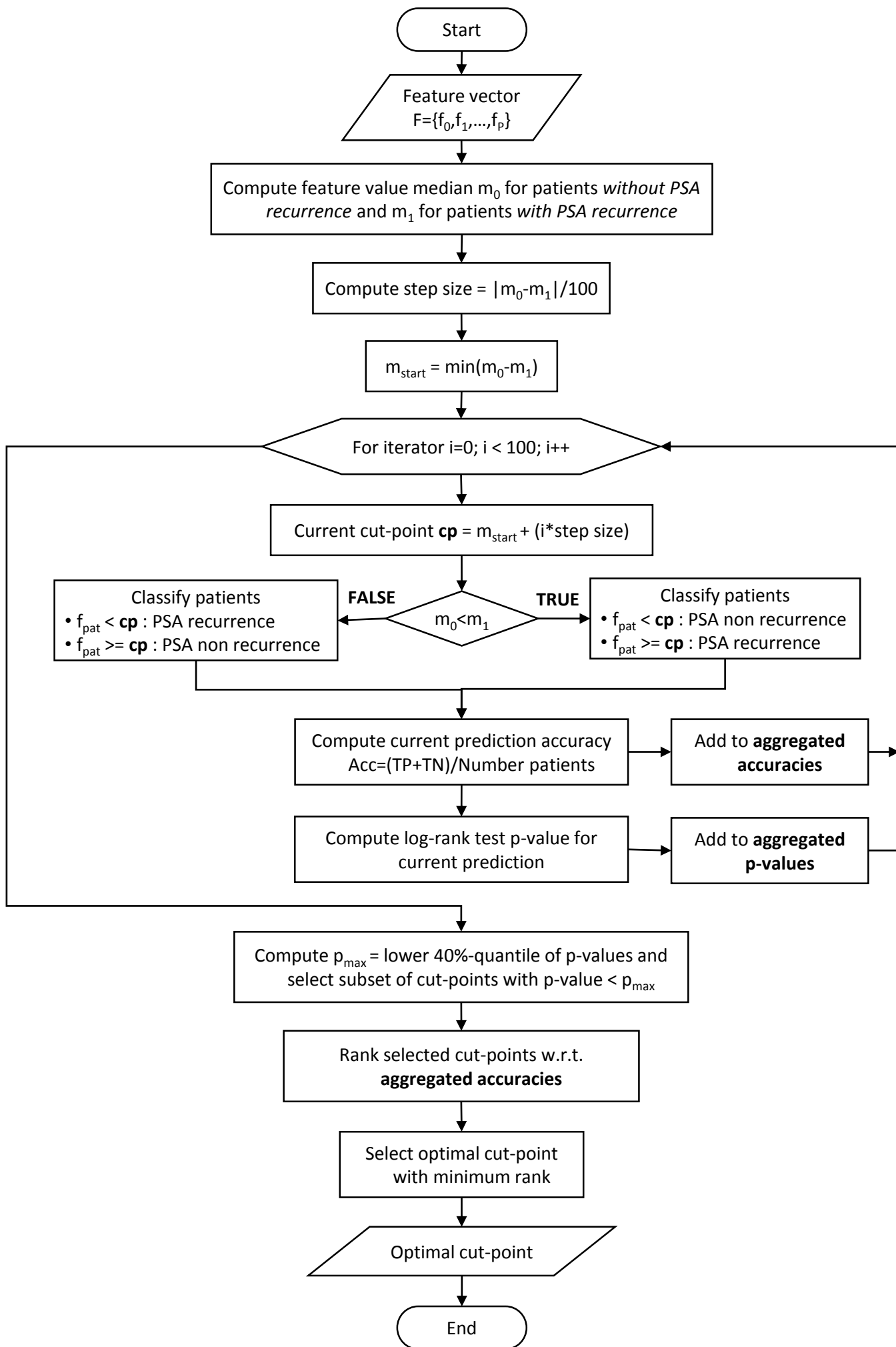


¹ROIs: Regions-of-interest, i.e., tumor region (non-intact glands), tumor micro environment (TME1, TME2), healthy (intact) glands, stroma

5) Feature ranking and phene discovery



5.1) Feature ranking and phene discovery: Cut-point selection



5.2) Feature ranking and phene discovery: CART feature selection

