Table 1. Summary of several results for lesion detection/segmentation on IDRiD dataset

Reference	Backbone	Loss	PR/%	SE/%	SP/%	ACC/%	AUPR/%	AUC/%	F1/%
		Не	morrhage	detection	/segmenta	ition			
Guo et al. (2019)	FCN	Top-k loss,	-	-	-	-	-	67.34	-
		Bin loss							
Yan et al.	U-Net	weighted	-	-	-	-	70.3	-	-
(2019a)		CE							
		Micro	oaneurysr	ns detecti	on/segme	ntation			
Sarhan et al.	FCN	Dice loss,	61.12	28.07	-	-	41.96	-	38.4877
(2019)(geometri		CE and	8						
c)		Triplet loss							
Guo et al. (2019)	FCN	Top-k loss,	-	-	-	-	-	46.27	-
		Bin loss							
Yan et al.	U-Net	weighted	_	-	_	-	52.5	-	-
(2019a)		CE							
Xue et al. (2019)	Mask-	log loss,	-	76.4	99.8	99.7	-	-	-
	RCNN	regression							
		loss, CE loss							
		Haı	d exudate	e detection	n/segment	ation			
Guo et al.	HED	Top-k loss,	-	95.74	-	-	-	98.71	95.57
(2020a)		Bin loss							
Guo et al. (2019)	FCN	Top-k loss,	-	-	_	-	-	79.45	-
		Bin loss							
Yan et al.	U-Net	weighted	-	-	_	-	88.9	-	-
(2019a)		CE							
Xue et al. (2019)	Mask-	log loss,	-	77.9	99.6	99.2	-	-	-
, ,	RCNN	regression							
		loss, CE loss							
		So	ft exudate	detection	/segmenta	ıtion			
Guo et al. (2019)	FCN	Top-k loss,	-	-	-	-	-	71.13	-
		Bin loss							
Yan et al.	U-Net	weighted	_	_	_	-	67.9	-	-
(2019a)		CE							

Table 2. Summary of several results for lesion detection/segmentation on E-ophtha dataset

Reference	Task	Backbone	Loss		PR/%	SE/%	SP/%	ACC/%	AUPR/%	AUC/%	F1/%
Carson et	MA	CNN	-		-	-	-	-	86	94	-
al. (2018)	classification										
Guo et al.	MA	FCN	Top-k	loss,	-	-	-	-	-	16.87	-
(2019)	segmentation		Bin los	s							
Xue et al.	MA	Mask-	log	loss,	-	67.2	99.8	99.7	-	-	-
(2019)	segmentation	RCNN	regress	ion							

loss and CE

loss

Carson et	Exudates	CNN	-	•	-	-	-	-	64	95	-
al. (2018)	classification										
Guo et al.	EX detection	HED	Top-k	loss,	-	86.44	-	-	-	91.84	87.01
(2020a)			Bin los	S							
Guo et al.	EX	FCN	Top-k	loss,	-	-	-	-	-	41.71	-
(2019)	segmentation		Bin los	s							
Xue et al.	EX	Mask-	log	loss,	-	84.6	98.8	98.4	-	-	-
(2019)	segmentation	RCNN	regress	ion							
			loss an	d CE							
			loss								
Playout et	Bright	U-Net	loss bas	sed on	78.50	80.02	99.88	99.77	-	-	79.25
al. (2019)	Lesion		Cohen'	s							
	segmentation		coeffici	ient							
Playout et	Red Lesion	U-Net	loss bas	sed on	75.26	75.62	99.99	99.88	-	-	75.44
al. (2019)	segmentation		Cohen'	s							
			coeffici	ent							

Table 3. Summary of several results for lesion detection/segmentation on DiaretDB1 dataset

Reference	Task	Backbone	Loss			PR/%	SE/%	SP/%	ACC/%	AUC/%	F1/%
Dai et al.	MA	CNN	-			99.7	87.8	-	96.1	93.4	-
(2018)	detection										
Adem	Exudate	CNN	-			-	99.2	97.97	-	-	-
(2018)	detection										
Playout et	Bright lesion	U-Net	loss	based	on	-	75.35	99.86	-	-	-
al. (2018)	segmentation		Cohe	n's coeffic	eient						
Playout et	Bright lesion	U-Net	loss	based	on	81.70	88.29	99.93	99.89	-	84.87
al. (2019)	segmentation		Cohe	ı's coeffic	eient						
Playout et	Red lesion	U-Net	loss	based	on	-	66.91	99.82	-	-	-
al. (2018)	segmentation		Cohe	n's coeffic	eient						
Playout et	Red lesion	U-Net	loss	based	on	78.96	85.18	99.89	99.83	-	81.95
al. (2019)	segmentation		Cohe	ı's coeffic	eient						

Table 4. Summary of several results for lesion detection/segmentation on other datasets

Reference	Task	Dataset	Backbone	Loss		SE/%	SP/%	AUC/%	mAP/%
van Grinsven et al.	HE detection	Kaggle	CNN	CE		83.7	85.1	89.4	-
(2016)									
van Grinsven et al.	HE detection	Messidor	CNN	CE		91.9	91.4	97.2	-
(2016)									
Huang et al. (2020)	HE	private	CNN	MSE,	IoU,	-	-	-	52.20

	segmentation			GIoU				
Yan et al. (2018a)	Drusen	STARE,	Encoder-	-	92.02	97.30	-	-
	segmentation	DRIVE	decoder					
			Network					
Adem (2018)	Exudate	DiaretDB0	CNN	-	100	98.41	-	-
	detection							
Adem (2018)	Exudate	DrimDB	CNN	-	100	98.44	-	-
	detection							
Tan et al. (2017)	EX detection	CLEOPATRA	CNN	log-likelihood	87.58	98.73	-	-
				function				
Tan et al. (2017)	HE detection	CLEOPATRA	CNN	log-likelihood	62.57	98.93	-	-
				function				
Tan et al. (2017)	MA	CLEOPATRA	CNN	log-likelihood	46.06	97.99	-	-
	detection			function				
Guo et al. (2019)	EX	DDR	FCN	Top-k loss,	-	-	55.46	-
	segmentation			Bin loss				
Guo et al. (2019)	SE	DDR	FCN	Top-k loss,	-	-	26.48	-
	segmentation			Bin loss				
Guo et al. (2019)	HE	DDR	FCN	Top-k loss,	-	-	35.86	-
	segmentation			Bin loss				
Guo et al. (2019)	MA	DDR	FCN	Top-k loss,	-	-	10.52	-
	segmentation			Bin loss				

 ${\bf Table~5.~Summary~of~several~results~for~vessel~segmentation~on~DRIVE~dataset}$ 

Reference	Backbone	Loss	SE/%	SP/%	ACC/%	AUC/%	F1/%
Khalaf et al. (2016)	CNN	-	83.97	95.62	94.56	-	-
Liskowski and Krawiec	CNN	CE	91.60	92.41	92.30	97.38	-
(2016)							
Yu et al. (2020)	CNN	-	76.43	98.03	95.24	97.23	-
Fu et al. (2016)	FCN	CBCE	76.03	-	95.23	-	-
Dasgupta and Singh	FCN	CE	76.91	98.01	95.33	97.44	-
(2017)							
Feng et al. (2017)	FCN	CBCE	78.11	98.39	95.60	97.92	-
Oliveira et al. (2018)	FCN	categorical CE	80.39	98.04	95.76	98.21	-
Zhang and Chung (2018)	U-Net	CE	87.23	96.18	95.04	97.99	-
He et al. (2018)	U-Net	Focal loss	77.61	97.92	95.19	-	81.29
Yan et al. (2018b)	U-Net	Proposed	76.53	98.18	95.42	97.52	-
		segment-level loss					
Yan et al. (2019b)	U-Net	CE	76.31	98.20	95.38	97.50	-
Wu et al. (2018)	U-Net	CE	78.44	98.19	95.67	98.07	-
Wu et al. (2020)	U-Net	CE	79.96	98.13	95.82	98.30	-
Wang et al. (2020)	U-Net	CE	78.49	98.13	95.67	97.88	82.41

Hu et al. (2018)	FCN	improved CE	77.72	97.93	95.33	97.59	-
Wu et al. (2019)	U-Net	CE	80.38	98.02	95.78	98.21	-
Soomro et al. (2019)	SegNet	CBCE	87	98.5	95.6	98.6	-
Zhang et al. (2019a)	U-Net	-	81.00	98.48	96.92	98.56	-
Wang et al. (2019a)	U-Net	CE and Jaccard	79.40	98.16	95.67	97.72	82.70
_		loss					
Ma et al. (2019)	U-Net	CE	79.16	98.11	95.70	98.10	-
Zhao et al. (2020a)	Dense U-	global pixel loss,	83.29	97.67	-	-	82.29
_	Net	local matting loss					
Mishra et al. (2020)	U-Net	CE	89.16	96.01	95.40	97.24	-
Feng et al. (2020)	FCN	MSE	76.25	98.09	95.28	96.78	-
Cherukuri et al. (2020)	Residual	MSE	84.25	98.49	97.23	98.70	-
	FCN						
Kromm and Rohr (2020)	CapsNet	margin loss	76.51	98.18	95.47	97.50	-
Liu et al. (2019a)	No-	MSE	80.72	97.80	95.59	97.79	82.25
	reference						
	net						
Wang et al.	U-Net		81.07	98.45	96.81	98.17	-
(MICCAI2020)							
Xu et al.(MICCAI2020)	U-Net	CE	91.2	94.7	-	98.1	-
Zhang et al.	U-Net	-	82.15	98.45	97.01	98.67	82.67
(MICCAI2020)							

Table 6. Summary of several results for vessel segmentation on STARE dataset

Reference	Backbone	Loss	SE/%	SP/%	ACC/%	AUC/%	F1/%
Liskowski and Krawiec	CNN	CE	93.07	93.04	93.09	98.20	-
(2016)							
Yu et al. (2020)	CNN	-	78.37	98.22	96.13	97.87	-
Fu et al. (2016)	FCN	CBCE	74.12	-	95.85	-	-
Oliveira et al. (2018)	FCN	categorical CE	83.15	98.58	96.94	99.05	-
Zhang and Chung (2018)	U-Net	CE	76.73	99.01	97.12	98.82	-
He et al. (2018)	U-Net	Focal loss	81.20	98.95	97.04	-	85.53
Yan et al. (2018b)	U-Net	Proposed segment-level	75.81	98.46	96.12	98.01	-
		loss					
Yan et al. (2019b)	U-Net	CE	77.35	98.57	96.38	98.33	-
Wu et al. (2020)	U-Net	CE	79.63	98.63	96.72	98.75	-
Wang et al. (2020)	U-Net	CE	90.24	99.34	98.49	99.60	91.84
Hu et al. (2018)	FCN	improved CE	75.43	98.14	96.32	97.51	-
Feng et al. (2020)	FCN	MSE	77.09	98.48	96.33	97	-
Soomro et al. (2019)	SegNet	CBCE	84.8	98.6	96.8	98.8	-
Cherukuri et al. (2020)	Residual	MSE	86.64	98.95	98.03	99.35	-
	FCN						

Zhao et al. (2020a)	Dense U- Net	global pixel loss, local matting loss	84.33	98.57	-	-	83.51
Mishra et al. (2020)	U-Net	CE	87.71	96.34	95.71	97.42	-
Liu et al. (2019a)	No- reference	MSE	77.71	98.43	96.23	97.93	80.36
	net						
Xu et al.(MICCAI2020)	U-Net	CE	92.8	95.4	-	98.6	-

Table 7. Summary	of covere	I regulte for vecce	coamontation on	CHASE DRI	datacat
Table 7. Summary	oi severa	i results for vesse	i segmentanon on	CHASE DDI	uataset

Reference	Backbone	Loss	SE/%	SP/%	ACC/%	AUC/%	F1/%
Fu et al. (2016)	FCN	CBCE	71.30	-	94.89	-	-
Oliveira et al. (2018)	FCN	categorical CE	77.79	98.64	96.53	98.55	-
Zhang and Chung (2018)	U-Net	CE	76.70	99.09	97.70	99.00	-
Yan et al. (2018b)	U-Net	Proposed segment-level loss	76.33	98.09	96.10	97.81	-
Yan et al. (2019b)	U-Net	CE	76.41	98.06	96.07	97.76	-
Wu et al. (2018)	U-Net	CE	75.38	98.47	96.37	98.25	-
Wu et al. (2020)	U-Net	CE	80.03	98.80	96.88	98.94	-
Wang et al. (2020)	U-Net	CE	79.48	98.42	96.48	98.47	82.20
Wu et al. (2019)	U-Net	CE	81.32	98.14	96.61	98.60	-
Soomro et al. (2019)	SegNet	CBCE	88.6	98.2	97.6	98.5	-
Zhang et al. (2019a)	U-Net	-	81.86	98.48	97.43	98.63	-
Cherukuri et al. (2020)	Residual FCN	MSE	80.17	99.08	97.88	98.64	-
Wang et al. (2019a)	U-Net	CE and Jaccard loss	80.74	98.21	96.61	98.12	80.37
Mishra et al. (2020)	U-Net	CE	88.05	96.51	96.01	97.63	-
Liu et al. (2019a)	No-	MSE	87.69	98.43	97.42	99.05	85.98
	reference						
	net						
Wang et al. (MICCAI2020)	U-Net	-	80.69	98.36	97.26	98.33	-
Xu et al. (MICCAI2020)	U-Net	CE	92.3	95.4	-	98.5	-

Table 8. Summary of several results for vessel segmentation on HRF dataset

Reference	Backbone	Loss	SE/%	SP/%	ACC/%	AUC/%	F1/%
Soomro et al. (2019)	SegNet	CBCE	82.9	96.1	96.2	98.5	-
Zhao et al. (2020a)	Dense U-	global pixel loss, local	78.09	98.18	-	-	78.13
	Net	matting loss					

Table 9. Summary of several results for OD/OC segmentation on Drishiti-GS dataset

Reference	Backbone	Loss	C	D	O	С	δ
			Dice/%	IoU/%	Dice/%	IoU/%	•
Edupuganti et al. (2018)	FCN	weighted CE	-	69.58	-	81.22	-
Mohan et al. (2018)	FCN	bootstrapped CE and Dice	96.4	-	-	-	-
		loss					
Mohan et al. (2019)	FCN	bootstrapped CE and Dice	97.13	-	-	-	-
		loss					
Liu et al. (2019e)	FCN	spatial-aware error	98	-	89	-	-
		function					
Shankaranarayana et	Encoder-	multi-class CE	96.3	-	84.8	-	0.1045
al.(2019)	decoder net						
Shah et al. (2019) (PSBN)	U-Net	logarithmic dice loss	95	91	88	80	-
Shah et al.	U-Net	logarithmic dice loss	96	93	89	80	-
(2019)(WRoIM)							
Wang et al. (2019c)	Deeplab,	dice coefficient loss,	97.4	-	90.1	-	0.048
	GAN	smoothness loss and					
		adversarial loss					
Wang et al. (2019b)	DeeplabV3+,	CE, MSE, Adversarial loss	96.1	-	86.2	-	-
	GAN						

## Table 10. Summary of several results for OD/OC segmentation on ORIGA dataset

Reference	Backbone	Loss	(	DD	(	OC	R	im	δ
			A/%	Е	A/%	Е	A/%	Е	•
Liu et al. (2019e)	FCN	spatial-aware error function	-	0.059	-	0.208	-	0.215	-
Fu et al. (2018a)	U-Net	proposed multi-label loss	98.3	0.071	93.0	0.230	94.1	0.233	0.071
Shankaranarayana et al.(2019)	Encoder- decoder net	multi-class CE	97.4	0.051	92.8	0.212	-	-	0.067
Yin et al. (2019)	RPN	Multi-label CE	98.6	0.066	94.2	0.208	94.9	0.224	0.065
Jiang et al. (2020)	atrous CNN and RPN	Smooth L1 loss and BCE	-	0.063	-	0.209	-	-	0.068

## Table 11. Summary of several results for OD/OC segmentation on RIM-ONE-r3 dataset

Reference	Backbone	Loss	OD					OC		δ	
			A/%	Е	Dice/%	IoU/%	A/%	Е	Dice/%	IoU/%	_
Shankaranaraya	Encoder-	multi-class CE	97.5	0.058	97.0	-	92.0	0.284	87.6	-	0.066
naet al. (2019)	decoder net										
Shah et al.	U-Net	logarithmic dice	-	-	91	84	-	-	75	60	-
(2019)(PSBN)		loss									
Shah et al.	U-Net	logarithmic dice			94	90	-	-	82	71	-
(2019)(WRoIM)		loss									
Wang et	Deeplab,	dice coefficient	-	-	96.8	-	-	-	85.6	-	0.049

al.(2019c)	GAN	loss, sn	noothness								
		loss, a	dversarial								
		loss									
Wang et al.	DeeplabV3	CE,	MSE, -	-	89.8	-	-	-	81.0	-	-
(2019b)	+, GAN	Advers	Adversarial loss								

Table 12. Summary of several results for OD/OC segmentation on REFUGE dataset

Reference	Backbone	Loss		OD			OC		R	lim	δ
			A/%	Е	Dice/%	A/%	Е	Dice/%	A/%	Е	
Wang et al.	RPN	Weighted CE, regression	-	-	95.3	-	-	87.2	-	-	0.047
(2019f)		loss									
Yin et al.	RPN	Multi-label CE	97.9	0.088	-	98.0	0.223	-	93.6	0.204	0.048
(2019)											
Wang et al.	Deeplab,	dice coefficient loss,	-	-	96.02	-	-	88.26	-	-	0.0450
(2019c)	GAN	smoothness loss and									
		adversarial loss									
Liu et al.	GAN	dice segmentation loss,	-	-	94.16	-	-	86.27	-	-	0.0481
(2019d)		adversarial loss and MSE									
		loss									

Table 13. Summary of several results for OD/OC segmentation on other datasets

Reference	Dataset	Backbone	Loss	(	OD	(	OC	δ
				Е	Dice/%	Е	Dice/%	•
Mohan et al.	DrionsDB	FCN	bootstrapped CE,	-	95.5	-	-	-
(2018)			Dice loss					
Mohan et al.	DrionsDB	FCN	bootstrapped CE,	-	96.6	-	-	-
(2019)			Dice loss					
Mohan et al.	MESSIDOR	FCN	bootstrapped CE,	-	95.7	-	-	-
(2018)			Dice loss					
Mohan et al.	MESSIDOR	FCN	bootstrapped CE,	-	96.8	-	-	-
(2019)			Dice loss					
Jiang et al.	SCES	atrous CNN,	Smooth L1 loss,	0.063	-	0.209	-	0.068
(2020)		RPN	BCE					
Sedai et al.	EyePACS	VAE	negative KL-	-	-	-	-	0.80
(2017a)			divergence, BCE					

Table 14. Summary of several results for DR diagnosis/grading

Reference	Dataset	Category	Backbone	Loss	SE/%	SP/%	AUC/%	Kappa/%
David et al. (2016)	Messidor-2	4	CNN	-	96.8	87.0	98.0	-
Gulshan et al. (2016)	Messidor-2	2	Inception-v3	-	87.0	98.5	99.0	-
Gargeya and Leng(2017)	Messidor-2	2	CNN	2-class	93	87	94	-
				categorical CE				
Wang et al. (2017)	Messidor	5	CNN	-	-	-	95.7	-
Lin et al. (2018)	Messidor	5	CNN	-	-	-	96.8	-

Gulshan et al. (2016)	EyePACS	2	Inception-v3	-	90.3	98.1	99.1	-
Gargeya and Leng(2017)	EyePACS	2	CNN	2-class	94	98	97	-
				categorical CE				
Gargeya and Leng(2017)	E-Ophtha	2	CNN	2-class	90	94	95	-
				categorical CE				
Quellec et al. (2017)	E-Ophtha	2	CNN	-	-	-	94.9	-
Wang et al. (2017)	Kaggle	5	CNN	-	-	-	85.4	-
Lin et al. (2018)	Kaggle	5	CNN	-	-	-	-	85.9
Roy et al. (2017)	Kaggle	5	CNN	-	-	-	-	86
Yang et al. (2017)	Kaggle	4	CNN	-	-	-	95.90	-
Quellec et al. (2017)	Kaggle	2	CNN	-	-	-	95.5	-
Galdran et al.	Kaggle	5	CNN	proposed	-	-	-	78.71 ±
(MICCAI2020)								0.28
Liu et al.(MICCAI2020)	Kaggle	-	CNN+GCN	-	-	-	-	72.7
Gondal et al. (2017)	DiaretDB1	2	CNN	-	93.6	97.6	95.4	-
Foo et al. (2020)	SiDRP14-15	5(No DR	U-Net,	binary CE	-	-	78.56	-
		here)	VGG16					
Foo et al. (2020)	IDRiD	5(No DR	U-Net,	binary CE	-	-	99.00	-
		here)	VGG16					
Lin et al. (2018)	private	5	CNN	-	-	-	-	87.5
Krause et al. (2017)	private	5 (moderate	Inception-v4	-	97.1	92.3	98.6	84
		or worse DR						
		here)						
Li et al. (2018b)	private	2	Inception-v3	-	92.5	98.5	95.5	-
Zhang et al. (2019b)	private	2	CNN	CE	97.5	97.7	97.7	-
Zhang et al. (2019b)	private	4	CNN	CE	98.1	98.9	-	-
Gulshan et al. (2019)	hospital in	2	CNN	-	92.1	95.2	98.0	-
	Sankara							
Gulshan et al. (2019)	hospitals in	2	CNN	-	88.9	92.2	96.3	-
	Aravind							
Liu et al.(MICCAI2020)	APTOS2019	-	CNN+GCN	-	-	-	-	91.2

Table 15. Summary of several results for glaucoma diagnosis/grading

Reference	Dataset	Backbone	Loss	SE/%	SP/%	ACC/%	BACC/%	AUC/%
Li et al. (2019a)/Li	RIM-ONE	CNN	K-L divergence	84.8	85.5	85.2	-	91.6
et al. (2020b)			function and CE					
dos Santos Ferreira	RIM-ONE,	U-Net, CNN	-	100	100	100	-	100
et al. (2018)	DRISHTI-							
	GS							
Zhao et al. (2019d)	ORIGA	CNN	contrastive loss	-		-	-	92
			and hinge loss					
Liao et al. (2020)	ORIGA	CNN	-	-	-		-	88

Li et al. (2019a)	LAG	CNN	K-L divergence	95.4	95.2	95.3	-	97.5
			function and CE					
Li et al. (2020b)	LAG	CNN	K-L divergence	95.4	96.7	96.2	-	98.3
			function and CE					
Wu et al.	LAG	Teacher-student	BCE	98.72	94.75	96.04		99.51
(MICCAI2020)		net						
Pal et al. (2018)	DRIONS-	Encoder-decoder	Reconstruction	-	-	-	-	92.3
	DB	network	loss and CE					
Fu et al. (2018b)	SCES	U-Net, ResNet50	Dice coefficient	84.78	83.80	-	84.29	91.83
			loss and CE					
Fu et al. (2018b)	SINDI	U-Net, ResNet50	Dice coefficient	78.76	71.15	-	74.95	81.73
			loss and CE					
Raghavendra et	Private	CNN	-	98.00	98.30	98.13	-	-
al.(2018)								
Li et al. (2018a)	Private	Inception-v3	-	95.6	92.0	-	-	98.6
Phene et al. (2019)	Private	Inception-v3	-	-	-	-	-	94.5
Chai et al. (2018)	Private	FCN, CNN,	CE	92.33	90.90	91.51	-	-
		Faster-RCNN						
Liu et al. (2019c)	Private	ResNet	CE	96.2	97.7	-	-	99.6
	FIGD							
Wu et al.	RIGA	Teacher-student	BCE	96.03	91.42	93.29	-	98.29
(MICCAI2020)		net						
Yu et al.	DRISHTI	CNN	CE, KL loss	91.43	74.19	86.14	-	89.63
(MICCAI2020)								
Li et al.	SIGF	CNN+LSTM	-	85.7	80.6	80.7	-	87.0
(MICCAI2020)								

Table 16. Summary of several results for AMD diagnosis/grading

Reference	Dataset	Backbone	Loss	Category	SE/%	SP/%	ACC/%	AUC/%	Kappa/%
Burlina et al.	AREDS	CNN with	-	2(1vs.3,4)	93.4	95.6	95.0	-	-
(2016)		SVM							
Burlina et al.	AREDS	CNN with	-	2	-	-	88.4~91.6	94~96	-
(2017)		SVM							
Horta et al.	AREDS	CNN with	-	2	66.34	88.95	79.04	84.76	-
(2017)		RF							
Govindaiah	AREDS	CNN	-	2	-	-	92.5	-	-
et al. (2018)									
Govindaiah	AREDS	CNN	-	-	-	-	83	-	-
et al. (2018)									
Burlina et al.	AREDS	ResNet50	Regression	4	-	-	75.7	-	-
(2018)			loss						
Peng et al.	AREDS	Inception-	-	6	59.0	93.0	67.1	-	55.8
(2018)		v3							
Burlina et al.	AREDS	ResNet50	Regression	9	-	-	59.1	-	-

(2018)			loss						
Grassmann et al. (2018)	AREDS, KORA	CNN	weighted k metric	13	-	-	63.3	-	-
Tan et al. (2018)	Collected	CNN	-	2	96.43	93.75	95.45	-	-