

PPF Report:
Adversarial Learning
-Diabetic Retinopathy

MSCV capstone project -- internal

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Contents

PAST:

- Reimplemented Adversarial Data Augmentation
- Detection Application: Diabetic Retinopathy(DR) Detection
- Improving DR detection

PRESENT:

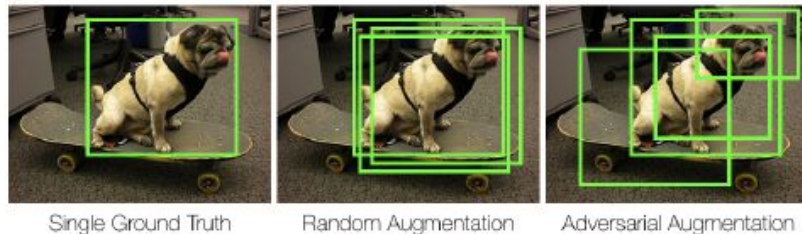
- DR Lesion Segmentation

FUTURE:

- Submit to IPMI

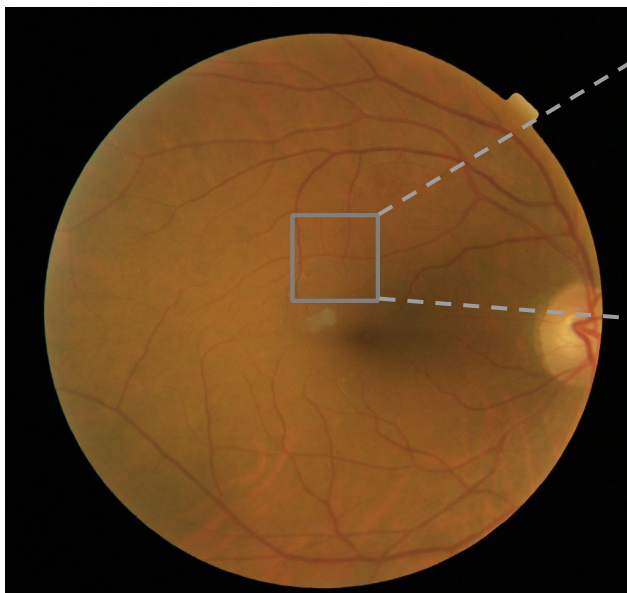
PAST: Reimplemented Adversarial Data Augmentation

- Introduce an adversarial function to generate (some distribution of) **maximally perturbed** version of the groundtruth which is hardest for the predictor to learn.
- First work to provide theoretic basis for data augmentation in terms of **an adversarial two player zero-sum game**.
 - predictor(maximize performance) vs constrained adversary(minimize expected performance).

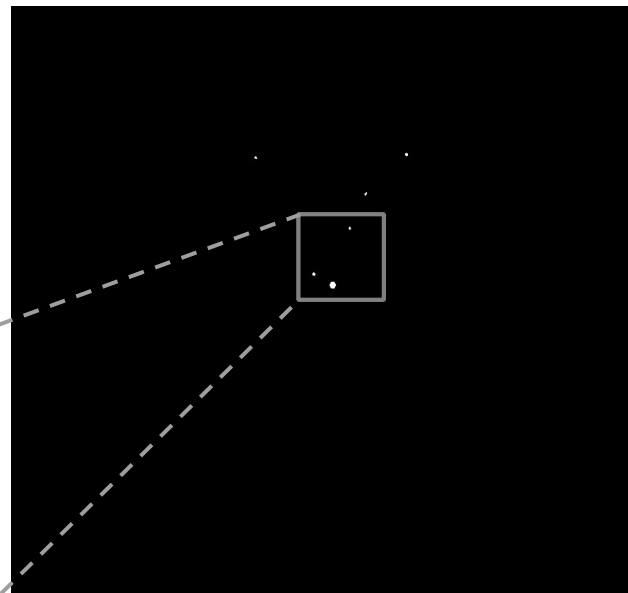
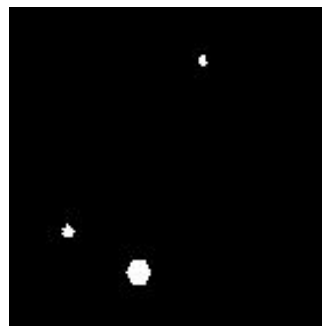
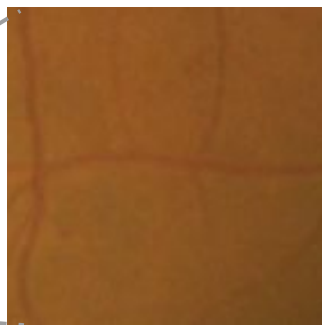


PAST: Diabetic Retinopathy(DR) Detection

- a worldwide leading cause of preventable blindness;
- affecting more than 25% of the estimated 425 million diabetic patients;



retinal image example

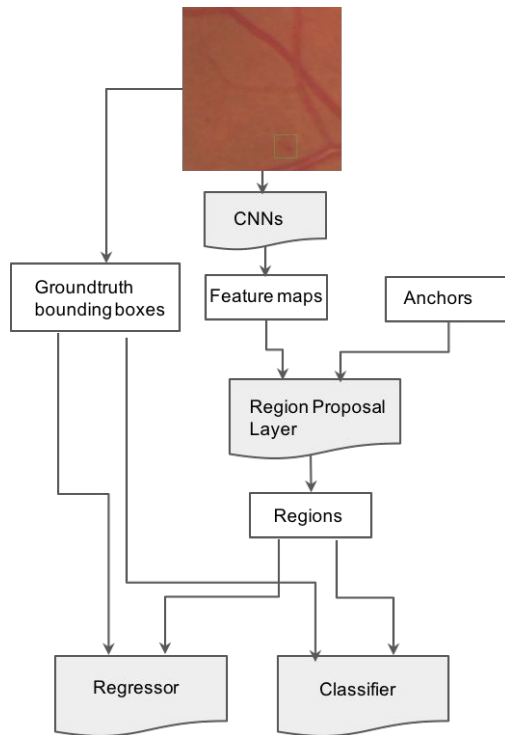


segmentation labels

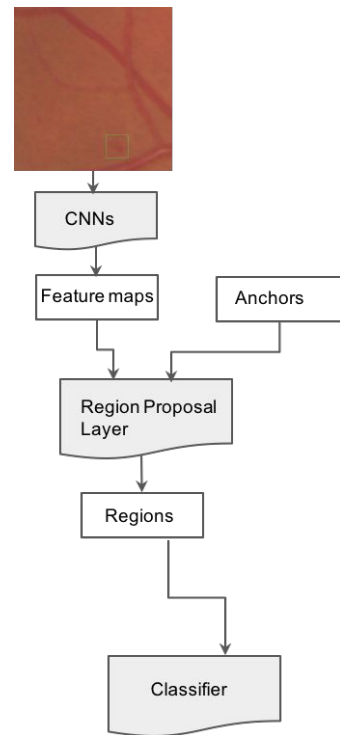
PAST: Improving DR detection

Inclusion of negative samples

- *positive samples*: patches containing at least one microaneurysm;
- *negative samples*: patches without microaneurysms;



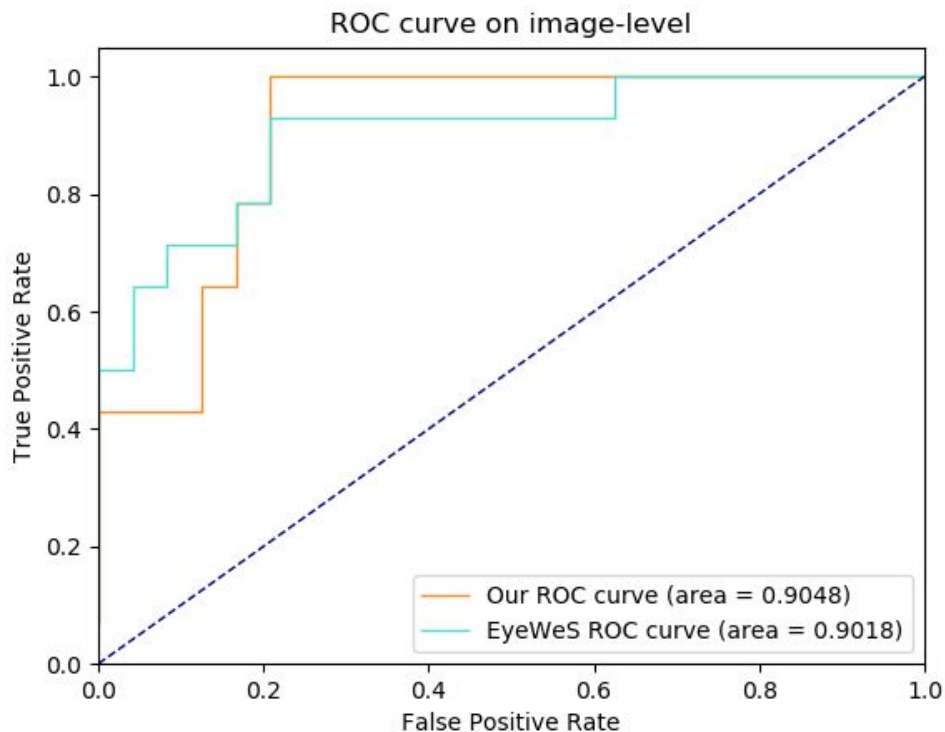
Faster RCNN, can only train on positive samples



Modified Faster RCNN, allow training on negative samples

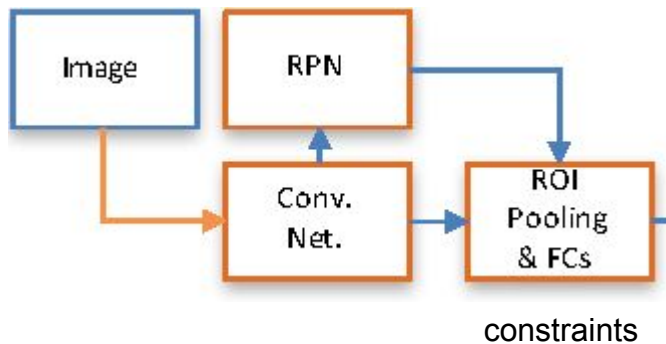
PAST: Improving DR Detection

Compare the approach to the already implemented baseline?



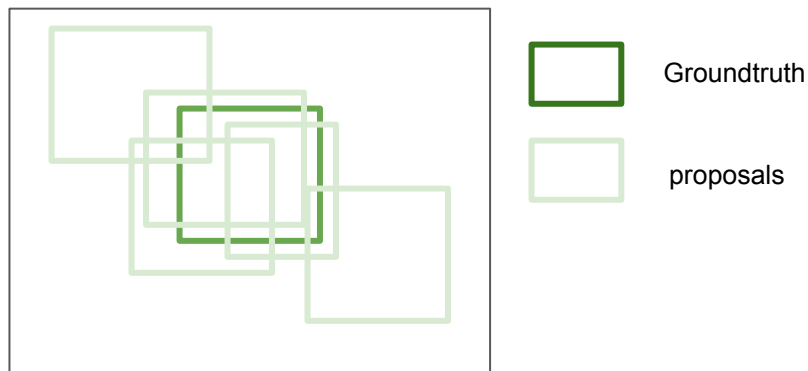
- Both with VGG16;
- Our method can simply transform the bounding box results into image scores (by summing up all the bounding-box scores, and then normalize all the scores to $[0, 1]$);
- Our method does not directly train for image-level classification.

PAST PLAN: Applying ADA to DR detection



$$\mathbf{G} = \begin{matrix} y_1 & y_2 & y_3 \\ \begin{matrix} y_1 \\ y_2 \\ y_3 \end{matrix} \end{matrix} \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix} + \begin{matrix} \phi(y_1) - \phi(y^*) & \phi(y_2) - \phi(y^*) & \phi(y_3) - \phi(y^*) \\ \phi(y_1) - \phi(y^*) & \phi(y_2) - \phi(y^*) & \phi(y_3) - \phi(y^*) \\ \phi(y_1) - \phi(y^*) & \phi(y_2) - \phi(y^*) & \phi(y_3) - \phi(y^*) \end{matrix} \begin{bmatrix} 0 & 0 & -1 \\ 0 & 0 & -1 \\ 0 & 0 & -1 \end{bmatrix}$$

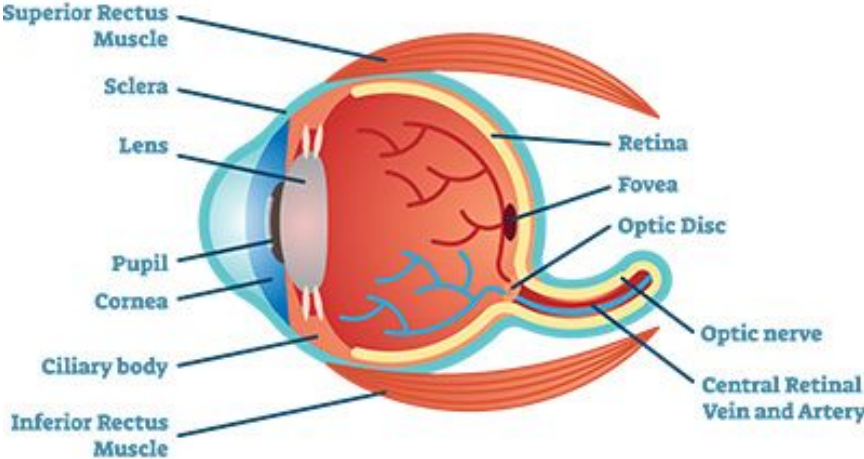
\mathbf{G}_ℓ \mathbf{G}_ϕ



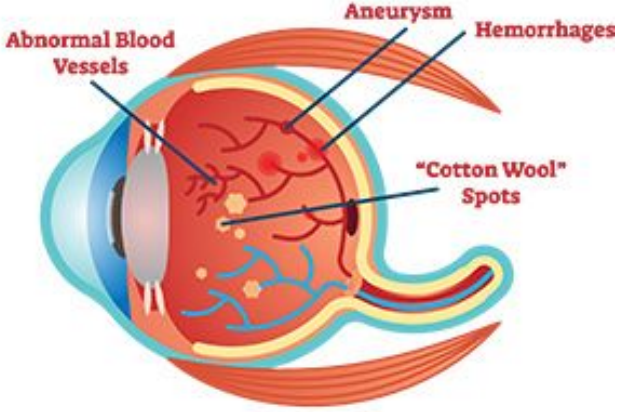
- ADA's Edge Box \rightarrow Faster RCNN's RPN
- BCE+Regression Loss \rightarrow IOU Loss

Not end-to-end

Diabetic Retionopathy



Healthy Eye



Diabetic Eye

Diabetic Retionpathy

Disease Detection/Grading: Classification of fundus images according to the severity level of diabetic retinopathy and diabetic macular edema.

Lesion Segmentation: Segmentation of retinal lesions associated with diabetic retinopathy as microaneurysms, hemorrhages, hard exudates and soft exudates.

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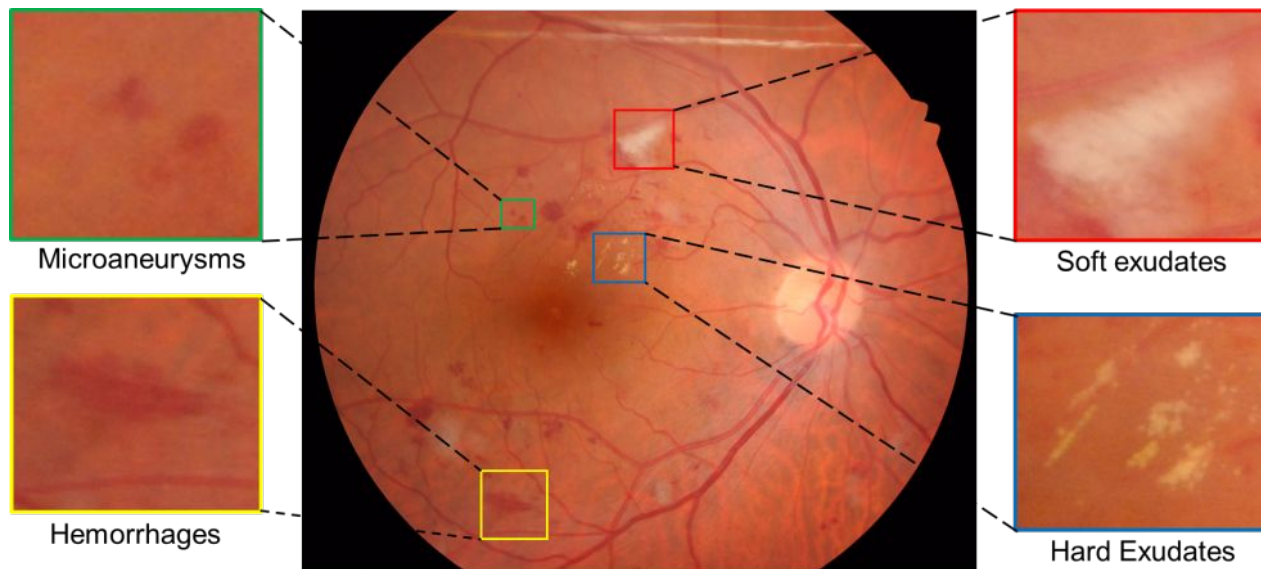
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DR Lesion Segmentation

Segmentation of retinal lesions associated with diabetic retinopathy as microaneurysms, hemorrhages, hard exudates and soft exudates.

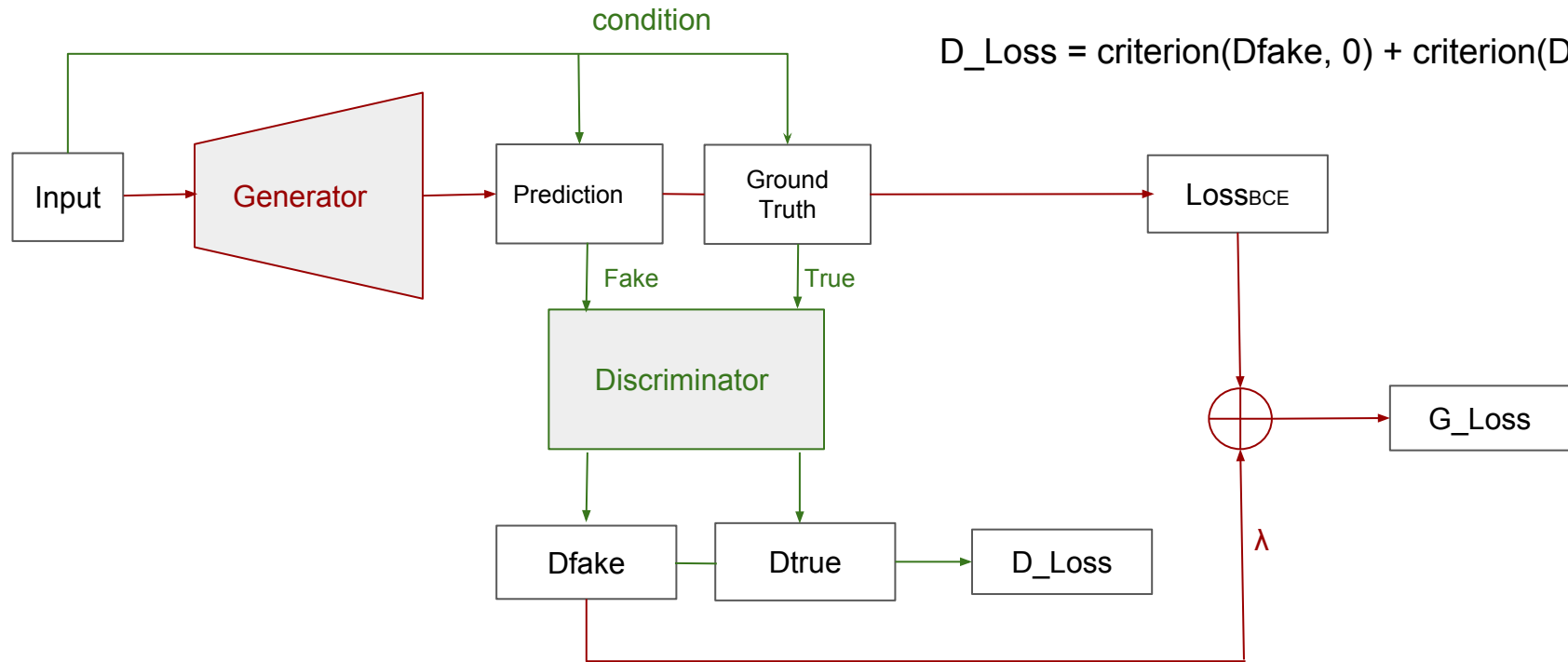


DR lesion segmentation dataset

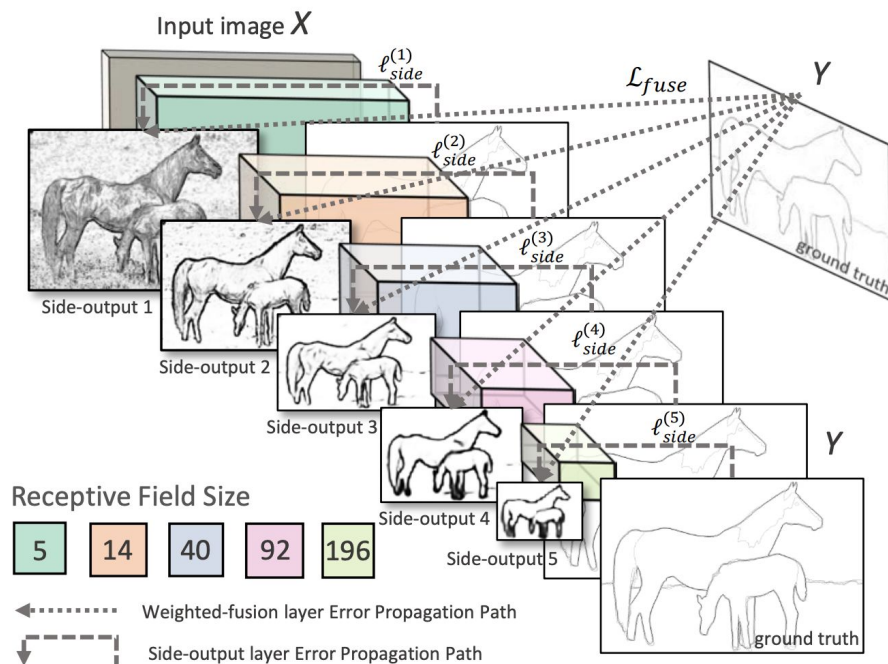
- [Indian Diabetic Retinopathy Image Dataset \(IDRiD\)](#)

	TRAINING SET	TESTING SET
microaneurysms (MA)	54	27
soft exudates (SE)	26	14
hard exudates (EX)	54	27
hemorrhages (HE)	53	27

Framework



Generator: HedNet Based



- **Loss function:**
Pixel-wise binary cross-entropy loss
- **Loss weights:**
Positive:Negative = 10:1

- [Xie, Saining, and Zhuowen Tu. "Holistically-nested edge detection." *Proceedings of the IEEE international conference on computer vision*. 2015.](#)

Performance Evaluation

Computing area under Precision (Positive Predictive Value) and Sensitivity (Recall) curve obtained by thresholding the results at 11 equally spaced instances. i.e. $[0, 0.1, \dots, 1]$.

The area under precision-recall (AUPR) is used to obtain a single score.

Average Precision

Quantitative Results(Average Precision)

Our Results

	EX	HE	MA	SE
Hednet	0.7935	0.4302	0.4270	0.5291
$\lambda = 0.01$	0.8183	0.4330	0.4468	-

LeaderBoard Results

Team Name	MA Score	RANK	HE Score	RANK	SE Score	RANK	EX Score	RANK
VRT	0.4951	2	0.6804	1	0.6995	1	0.7127	11
PATech	0.474	3	0.649	2	-	-	0.885	1
iFLYTEK-MIG	0.5017	1	0.5588	3	0.6588	3	0.8741	2
SOONER	0.4003	5	0.5395	4	0.5369	7	0.739	10
SAIHST	-	-	-	-	-	-	0.8582	3
lzyuncc_fusion	-	-	-	-	0.6259	4	0.8202	4
SDNU	0.4111	4	0.4572	7	0.5374	6	0.5018	17
CIL	0.392	6	0.4886	5	0.5024	8	0.7554	8
MedLabs	0.3397	8	0.3705	8	0.2637	10	0.7863	5
AIMIA	0.3792	7	0.3283	10	0.2733	9	0.7662	6

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FUTURE: Milestones

