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GTA5 ⟶ Cityscapes																						
Method	**Feat. Ex.**	**Class IoU**																			**mIoU**	**Code**
		Road	Sidewalk	Building	Wall	Fence	Pole	T. Light	T. Sign	Veg.	Terrain	Sky	Person	Rider	Car	Truck	Bus	Train	M.bike	Bicycle		
FCN-ITW [1]	VGG-16	70.4	32.4	62.1	14.9	5.4	10.9	14.2	2.7	79.2	21.3	64.6	44.1	4.2	70.4	8.0	7.3	0.0	3.5	0.0	27.1	-
SIBAN [2]	VGG-16	83.4	13.0	77.8	20.4	17.5	24.6	22.8	9.6	81.3	29.6	77.3	42.7	10.9	76.0	22.8	17.9	5.7	14.2	2.0	34.2	-
CyCADA [3]	VGG-16	85.2	37.2	76.5	21.8	15.0	23.8	22.9	21.5	80.5	31.3	60.7	50.5	9.0	76.9	17.1	28.2	4.5	9.8	0.0	35.4	✓
CBST-SP [4]	VGG-16	90.4	50.8	72.0	18.3	9.5	27.2	28.6	14.1	82.4	25.1	70.8	42.6	14.5	76.9	5.9	12.5	1.2	14.0	28.6	36.1	-
DCAN [5]	VGG-16	82.3	26.7	77.4	23.7	20.5	20.4	30.3	15.9	80.9	25.4	69.5	52.6	11.1	79.6	24.9	21.2	1.3	17.0	6.7	36.2	✓
CLAN [6]	VGG-16	88.0	30.6	79.2	23.4	20.5	26.1	23.0	14.8	81.6	34.5	72.0	45.8	7.9	80.5	26.6	29.9	0.0	10.7	0.0	36.6	-
DPR [7]	VGG-16	87.3	35.7	79.5	32.0	14.5	21.5	24.8	13.7	80.4	32.0	70.5	50.5	16.9	81.0	20.8	28.1	4.1	15.5	4.1	37.5	-
MADAN [8]	VGG-16	86.2	37.7	79.1	20.1	17.8	15.5	14.5	21.4	78.5	-	73.4	49.7	16.8	77.8	-	28.3	-	17.7	27.5	41.4	✓
SIM [9]	VGG-16	88.1	35.8	83.1	25.8	23.9	29.2	28.8	28.6	83.0	36.7	82.3	53.7	22.8	82.3	26.4	38.6	0.0	19.6	17.1	42.4	✓
FDA-MBT [10]	VGG-16	86.1	35.1	80.6	30.8	20.4	27.5	30.0	26.0	82.1	30.3	73.6	52.5	21.7	81.7	24.0	30.5	29.9	14.6	24.0	42.2	✓
LTIR [11]	VGG-16	92.5	54.5	83.9	34.5	25.5	31.0	30.4	18.0	84.1	39.6	83.9	53.6	19.3	81.7	21.1	13.6	17.7	12.3	6.5	42.3	✓
TGCF-DA + SE [12]	VGG-16	90.2	51.5	81.1	15.0	10.7	37.5	35.2	28.9	84.1	32.7	75.9	62.7	19.9	82.6	22.9	28.3	0.0	23.0	25.4	42.5	-
DCAN [5]	ResNet-101	88.5	37.4	79.3	24.8	16.5	21.3	26.3	17.4	80.8	30.9	77.6	50.2	19.2	77.7	21.6	27.1	2.70	14.3	18.1	38.5	✓
DLOW [13]	ResNet-101	87.1	33.5	80.5	24.5	13.2	29.8	29.5	26.6	82.6	26.7	81.8	55.9	25.3	78.0	33.5	38.7	0.0	22.9	34.5	42.3	✓
AdaptSegNet [14]	ResNet-101	86.5	36.0	79.9	23.4	23.3	23.9	35.2	14.8	83.4	33.3	75.6	58.5	27.6	73.7	32.5	35.4	3.9	30.1	28.1	42.4	✓
SIBAN [2]	ResNet-101	88.5	35.4	79.5	26.3	24.3	28.5	32.5	18.3	81.2	40.0	76.5	58.1	25.8	82.6	30.3	34.4	3.4	21.6	21.5	42.6	-
CyCADA [3]	ResNet-101	86.7	35.6	80.1	19.8	17.5	38.0	39.9	41.5	82.7	27.9	73.6	64.9	19.0	65.0	12.0	28.6	4.5	31.1	42.0	42.7	✓
CLAN [6]	ResNet-101	87.0	27.1	79.6	27.3	23.3	28.3	35.5	24.2	83.6	27.4	74.2	58.6	28.0	76.2	33.1	36.7	6.7	31.9	31.4	43.2	-
SWD [15]	ResNet-101	92.0	46.4	82.4	24.8	24.0	35.1	33.4	34.2	83.6	30.4	80.9	56.9	21.9	82.0	24.4	28.7	6.1	25.0	33.6	44.5	✓
SSF-DAN [16]	ResNet-101	90.3	38.9	81.7	24.8	22.9	30.5	37.0	21.2	84.8	38.8	76.9	58.8	30.7	85.7	30.6	38.1	5.9	28.3	36.9	45.4	-
ADVENT [17]	ResNet-101	89.4	33.1	81.0	26.6	26.8	27.2	33.5	24.7	83.9	36.7	78.8	58.7	30.5	84.8	38.5	44.5	1.7	31.6	32.4	45.5	✓
IntraDA [18]	ResNet-101	90.6	37.1	82.6	30.1	19.1	29.5	32.4	20.6	85.7	40.5	79.7	58.7	31.1	86.3	31.1	86.3	0.0	30.2	35.8	46.3	✓
MSL [19]	ResNet-101	89.4	43.0	82.1	30.5	21.3	30.3	34.7	24.0	85.3	39.4	78.2	63.0	22.9	84.6	36.4	43.0	5.5	34.7	33.5	46.4	✓
PANDA [20]	ResNet-101	92.4	51.3	82.9	31.8	24.9	32.6	35.8	20.4	84.5	38.7	79.8	60.0	25.8	85.1	33.7	44.1	9.0	27.5	22.6	46.5	-
DPR [7]	ResNet-101	92.3	51.9	82.1	29.2	25.1	24.5	33.8	33.0	82.4	32.8	82.2	58.6	27.2	84.3	33.4	46.3	2.2	29.5	32.3	46.5	-
SWLS [21]	ResNet-101	92.7	48.0	78.8	25.7	27.2	36.0	42.2	45.3	80.6	14.6	66.0	62.1	30.4	86.2	28.0	45.6	35.9	16.8	34.7	47.2	-
PyCDA [22]	ResNet-101	90.5	36.3	84.4	32.4	28.7	34.6	36.4	31.5	86.8	37.9	78.5	62.3	21.5	85.6	27.9	34.8	18.0	22.9	49.3	47.4	-
LSE+FL [23]	ResNet-101	90.2	40.0	83.5	31.9	26.4	32.6	38.7	37.5	81.0	34.2	84.6	61.6	33.4	82.5	32.8	45.9	6.7	29.1	30.6	47.5	-
BDL [24]	ResNet-101	91.0	44.7	84.2	34.6	27.6	30.2	36.0	36.0	85.0	43.6	83.0	58.6	31.6	83.3	35.3	49.7	3.3	28.8	35.6	48.5	✓
CSCL [25]	ResNet-101	89.6	50.4	83.0	35.6	26.9	31.1	37.3	35.1	83.5	40.6	84.0	60.6	34.3	80.9	35.1	47.3	0.5	34.5	33.7	48.6	-
CrCDA [26]	ResNet-101	92.4	55.3	82.3	31.2	29.1	32.5	33.2	35.6	83.5	34.8	84.2	58.9	32.2	84.7	40.6	46.1	2.1	31.1	32.7	48.6	-
SISC-PWL [27]	ResNet-101	89.0	45.2	78.2	22.9	27.3	37.4	46.1	43.8	82.9	18.6	61.2	60.4	26.7	85.4	35.9	44.9	36.4	37.2	49.3	49.0	-
SIM [9]	ResNet-101	90.6	44.7	84.8	34.3	28.7	31.6	35.0	37.6	84.7	43.3	85.3	57.0	31.5	83.8	42.6	48.5	1.9	30.4	39.0	49.2	✓
LDR [28]	ResNet-101	90.8	41.4	84.7	35.1	27.5	31.2	38.0	32.8	85.6	42.1	84.9	59.6	34.4	85.0	42.8	52.7	3.4	30.9	38.1	49.5	-
CCM [29]	ResNet-101	93.5	57.6	84.6	39.3	24.1	25,2	35.0	17.3	85.0	40.6	86.5	58.7	28.7	85.8	49.0	56.4	5.4	31.9	43.2	49.9	✓
CAG-UDA [30]	ResNet-101	90.4	51.6	83.8	34.2	27.8	38.4	25.3	48.4	85.4	38.2	78.1	58.6	34.6	84.7	21.9	42.7	41.1	29.3	37.2	50.2	✓
LTIR [11]	ResNet-101	92.9	55.0	85.3	34.2	31.1	34.9	40.7	34.0	85.2	40.1	87.1	61.0	31.1	82.5	32.3	42.9	0.3	36.4	46.1	50.2	✓
FDA-MBT [10]	ResNet-101	92.5	53.3	82.4	26.5	27.6	36.4	40.6	38.9	82.3	39.8	78.0	62.6	34.4	84.9	34.1	53.1	16.9	27.7	46.4	50.5	✓
CRST(MRKLD) + TPLD [31]	ResNet-101	94.2	60.5	82.8	36.6	16.6	39.3	29.0	25.5	85.6	44.9	84.4	60.6	27.4	84.1	37.0	47.0	31.2	36.1	50.3	51.2	-
IAST [32]	ResNet-101	93.8	57.8	85.1	39.5	26.7	26.2	43.1	34.7	84.9	32.9	88.0	62.6	29.0	87.3	39.2	49.6	23.2	34.7	39.6	51.5	✓
IAST-MST [32]	ResNet-101	94.1	58.8	85.4	39.7	29.2	25.1	43.1	34.2	84.8	34.6	88.7	62.7	30.3	87.6	42.3	50.3	24.7	35.2	40.2	52.2	✓
CyCADA [3]	DRN-26	79.1	33.1	77.9	23.4	17.3	32.1	33.3	31.8	81.5	26.7	69.0	62.8	14.7	74.5	20.9	25.6	6.9	18.8	20.4	39.5	✓
CrDoCo [33]	DRN-26	95.1	49.2	86.4	35.2	22.1	36.1	40.9	29.1	85.0	33.1	75.8	67.3	26.8	88.9	23.4	19.3	4.3	25.3	13.5	45.1	(✓)
CBST-SP+MST [4]	ResNet-38	89.6	58.9	78.5	33.0	22.3	41.4	48.2	39.2	83.6	24.3	65.4	49.3	20.2	83.3	39.0	48.6	12.5	20.3	35.3	47.0	-