

# On-Device AI 실습: Knowledge Distillation

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Dongkun Shin  
Intelligent Embedded Systems Lab.  
Sungkyunkwan University

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## 1. Baseline 학습 (Cross-Entropy Loss)

- Teacher 모델과 Student 모델을 각각 Cross-Entropy Loss만으로 학습시켜 정확도를 비교합니다.

## 2. Knowledge Distillation (Soft Targets)

- Teacher의 softmax 출력을 활용한 Knowledge Distillation을 적용하고, temperature 및 loss weight에 따른 영향을 분석합니다.

## 3. Cosine Loss Minimization (Cosine Loss)

- Teacher와 Student의 convolutional feature를 추출하여, CosineEmbeddingLoss를 적용해 내부 표현 유사도를 증가시키는 방식으로 학습합니다.

## 4. Intermediate Regressor (Regressor + MSE)

- Teacher의 feature map과 Student의 regressed feature map을 MSE로 정렬하며, 중간 표현을 직접 학습합니다.

# Setup

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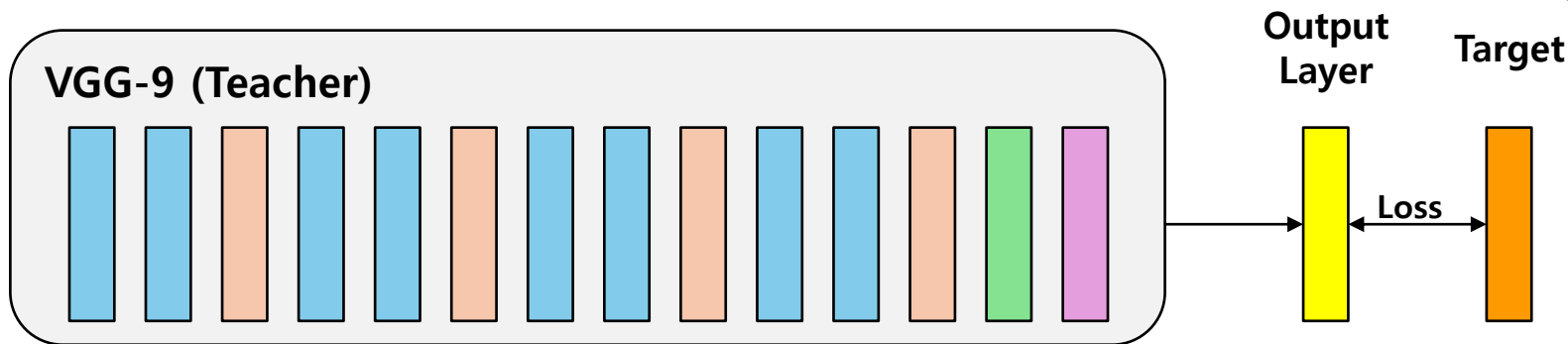
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- 실습 자료 "Knowledge Distillation.ipnyb"을 colab에서 실행해주세요
- Colab 런타임을 **GPU(T4)**로 설정해 주세요
- Setup 코드 셀을 실행해 필요한 패키지를 설치해주세요

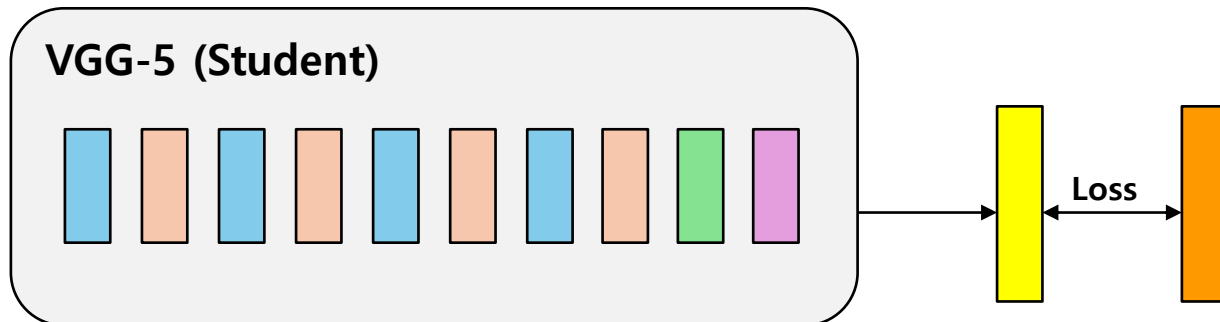
# Baseline 학습 (Cross-Entropy Loss)

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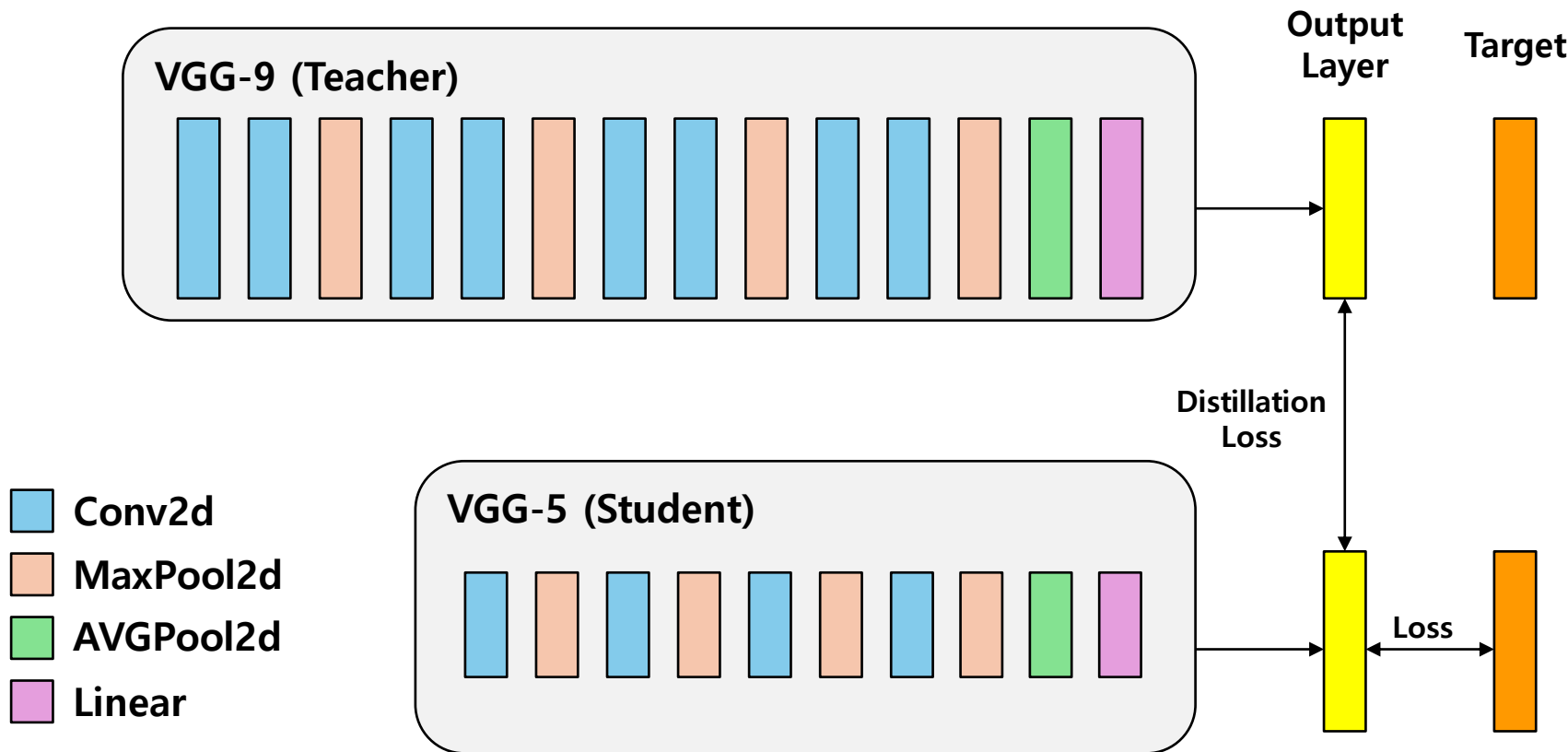
- Conv2d
- MaxPool2d
- AVGPoool2d
- Linear



# Knowledge Distillation (Soft Targets)

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- KL divergence:

$$D_{\text{KL}}(P \parallel Q) = \sum_{x \in \mathcal{X}} P(x) \log \frac{P(x)}{Q(x)}$$

- Distillation Loss:

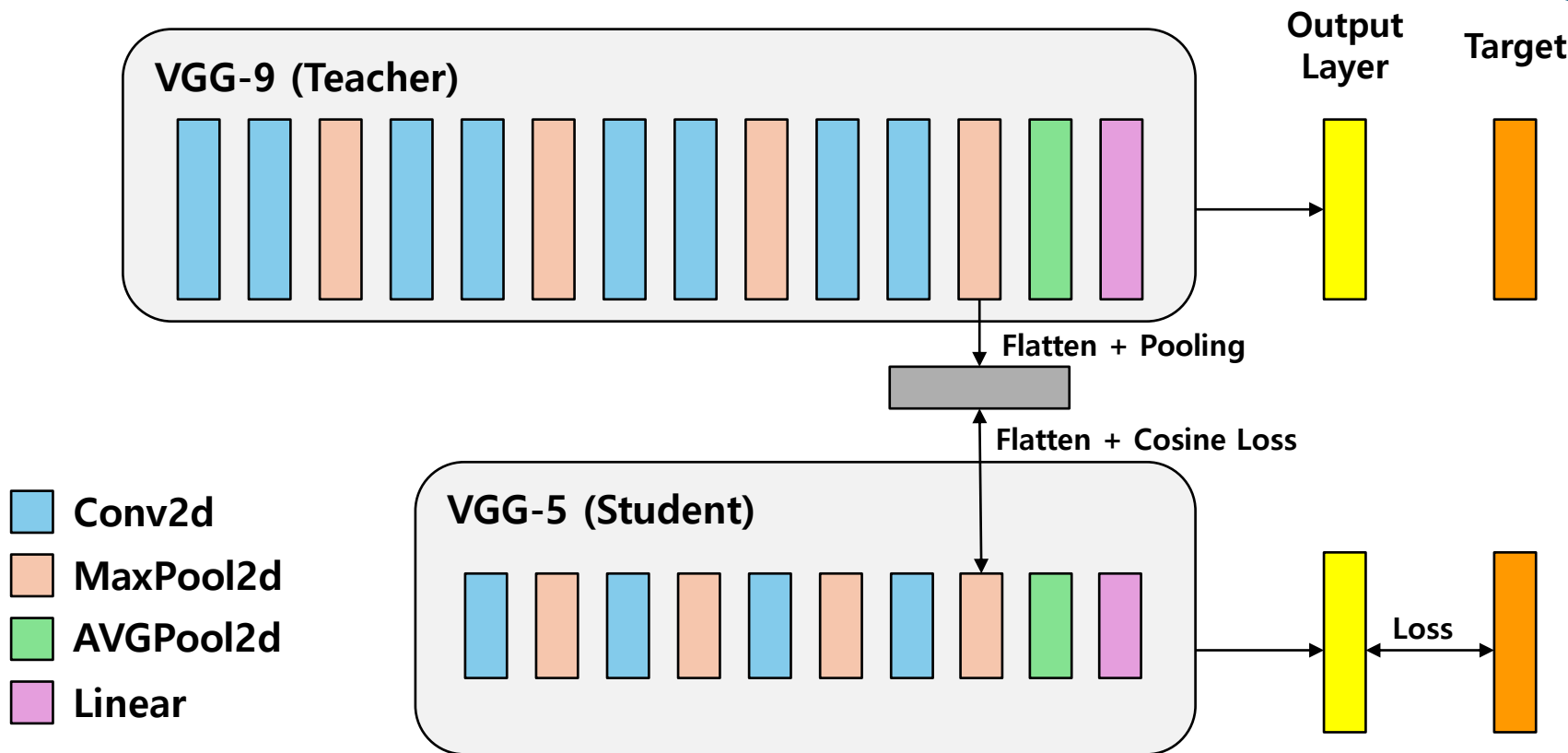
$$- \frac{D_{\text{KL}}(P \parallel Q)}{N} \times T^2$$

- N: batch size
- T: temperature (loss의 scale이  $T^2$ 에 반비례하므로  $T^2$ 을 곱해 loss를 보정)

# Cosine Loss Minimization (Cosine Loss)

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# Intermediate Regressor (Regressor + MSE)

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