



# 트랜스포머 NAS

성명 신익수

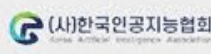
소속 한국전자통신연구원

**SUBJECT**

인공지능 기술의 대중화 (AI Democratization)를 위한  
TANGO 커뮤니티 3회 컨퍼런스

주관 ETRI ( 한국전자통신연구원 ) 주최 과학기술정보통신부 IITP 정보통신기획평가원

후원





## 목 차

1

NAS 기술 발전 History

01

2

연구 문제 설정

07

3

신규 NAS 기술

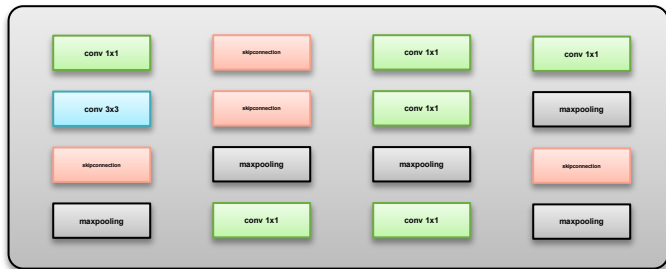
13

# NAS 기술 발전 History

## Classic NAS



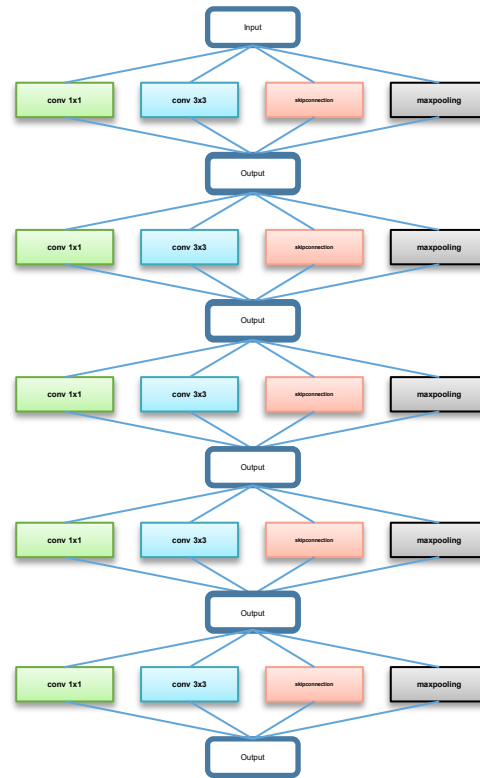
### Generation $N$



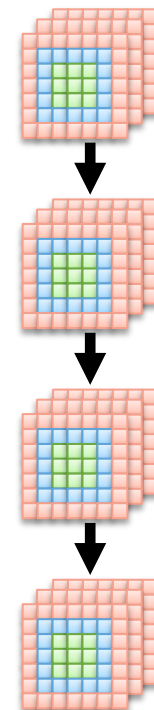
### Generation $N+1$



## OneShot NAS

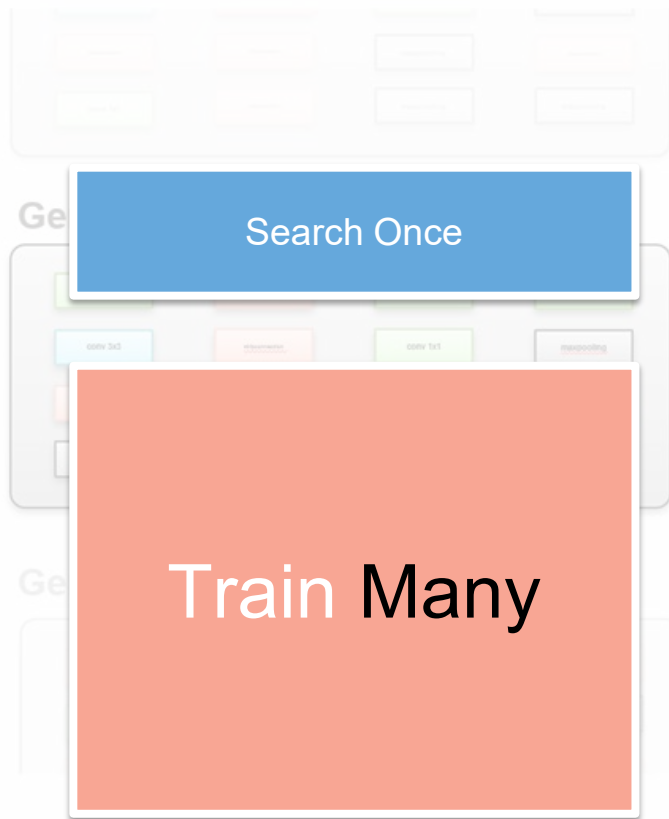


## Once-for-all Style OneShot NAS

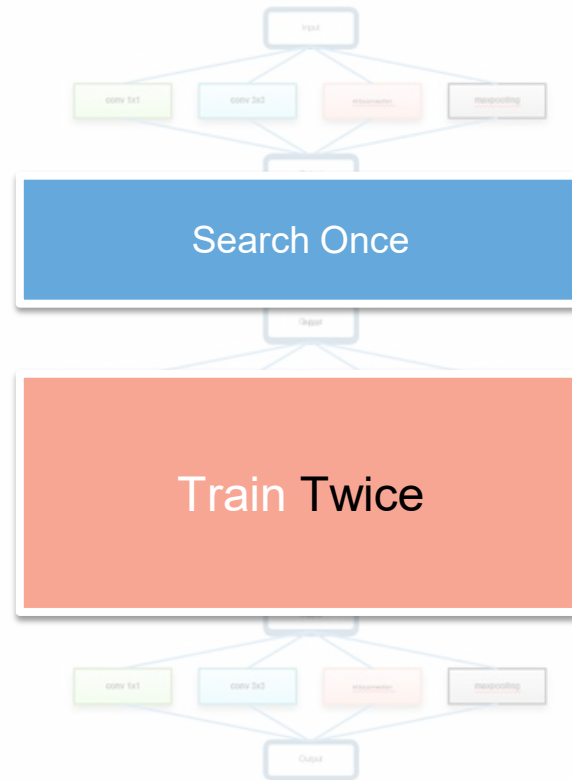


# NAS 기술 발전 History

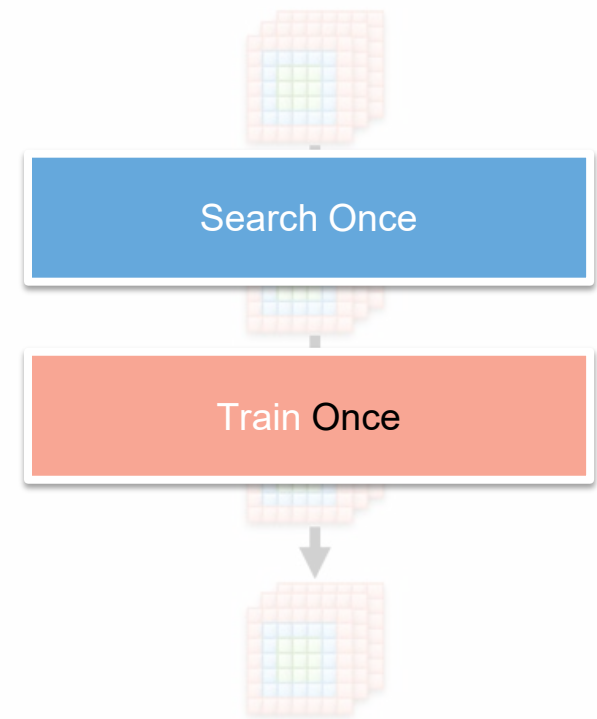
## Classic NAS



## OneShot NAS

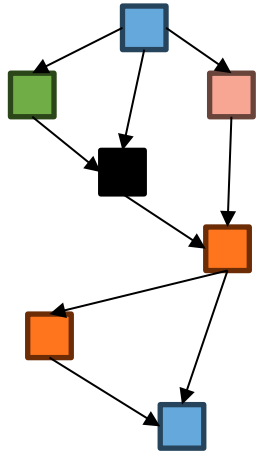


## Once-for-all Style OneShot NAS



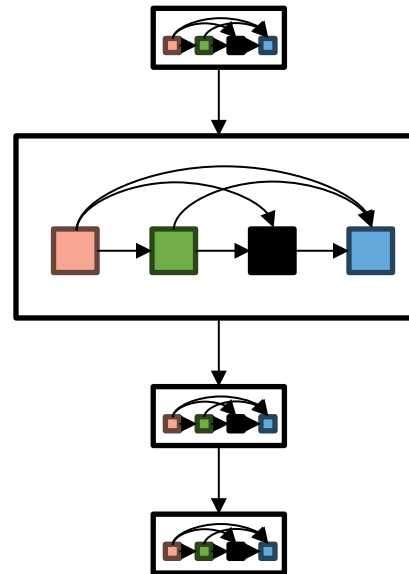
# NAS 기술 발전 History

## Classic NAS



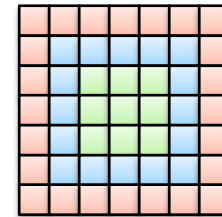
다양한 구조 생성 가능

## OneShot NAS



블록 단위 구조

## Once-for-all Style OneShot NAS



연산의 크기만 조절



# NAS 기술 발전 History

## Classic NAS

## OneShot NAS

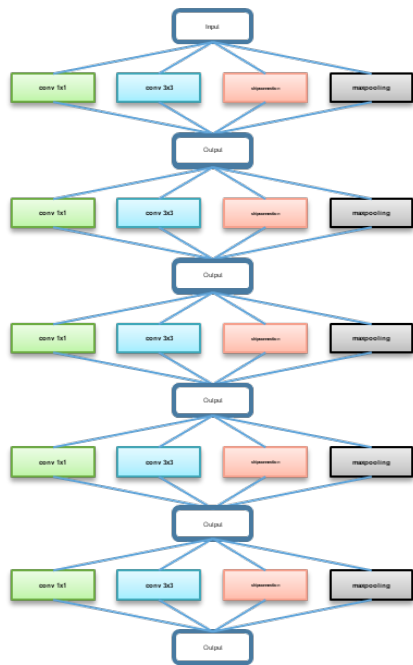
## Once-for-all Style OneShot NAS

탐색 속도 증가

구조 다양성 감소

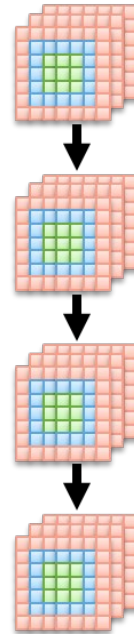
# 연구 문제 설정

## OneShot NAS



구조 다양성

## Once-for-all Style OneShot NAS



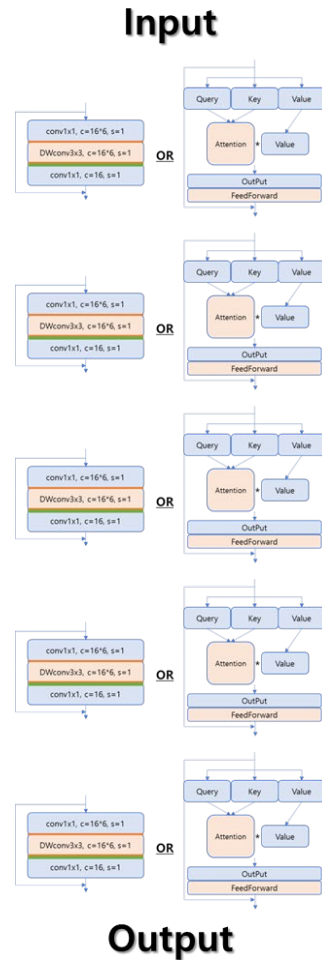
고속 탐색

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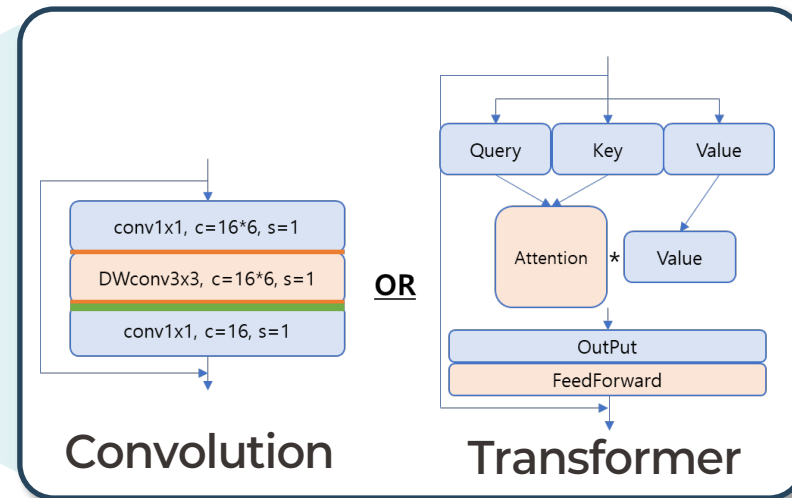
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# 연구 문제 설정



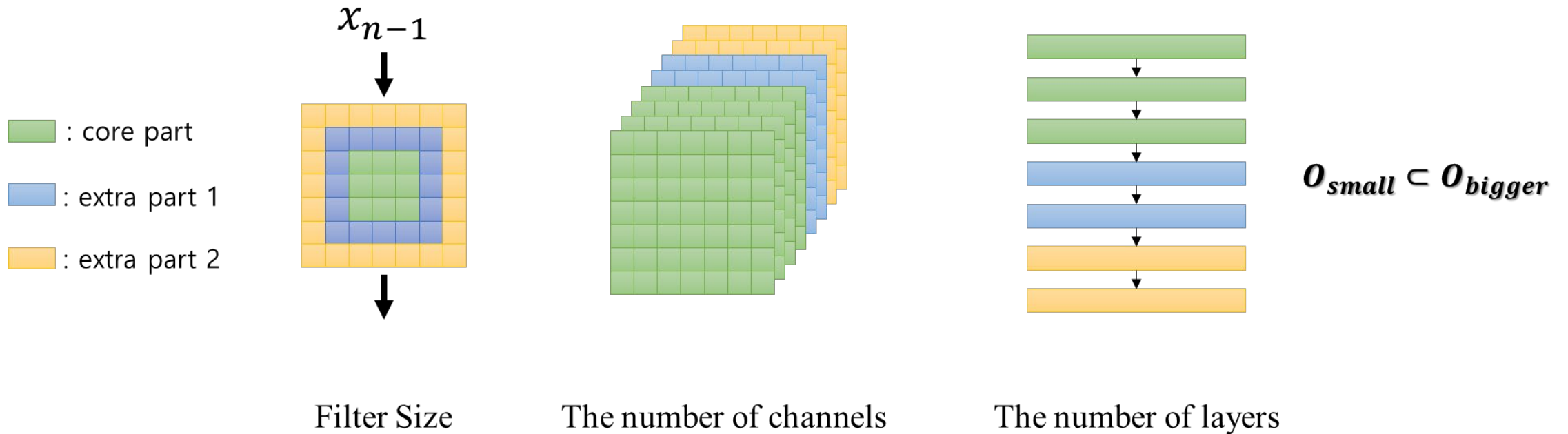
## 각 레이어마다 연산자 선택권





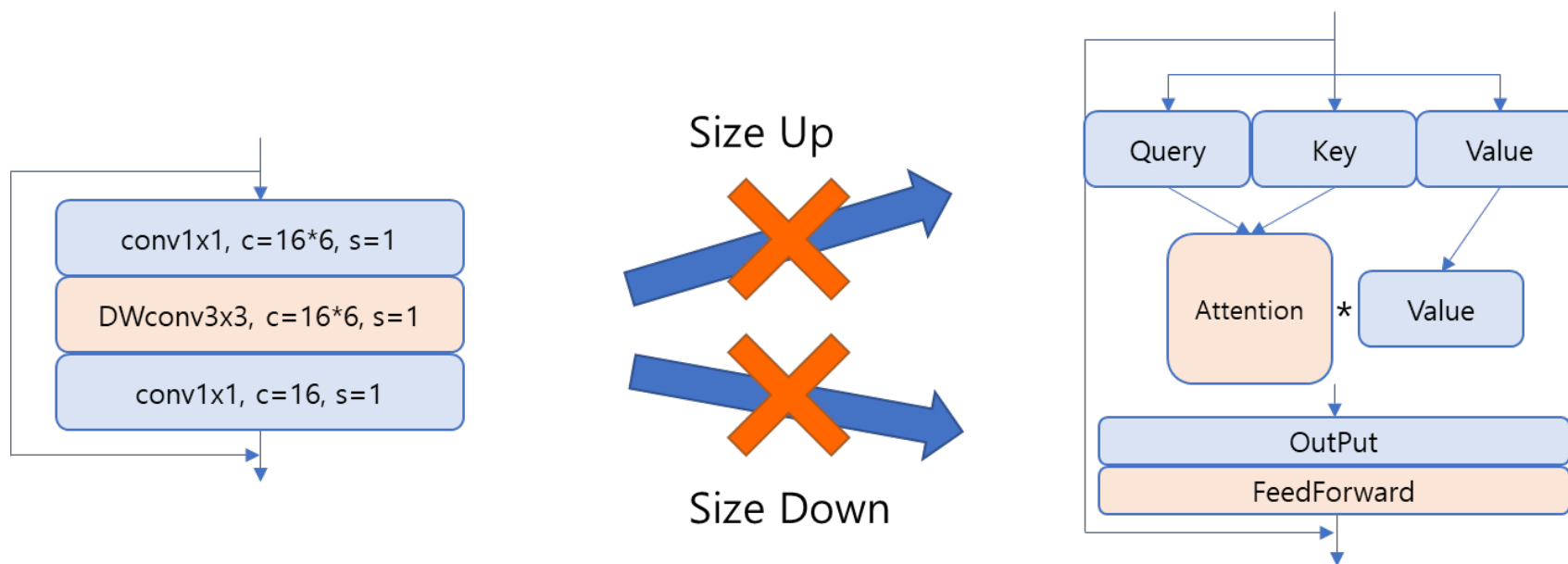
# 연구 문제 설정

## 고속 탐색의 비결은 연산자들 간의 가중치 공유 구조



# 연구 문제 설정

## 다른 종류의 연산자들 간에는 가중치 공유 구조 불가

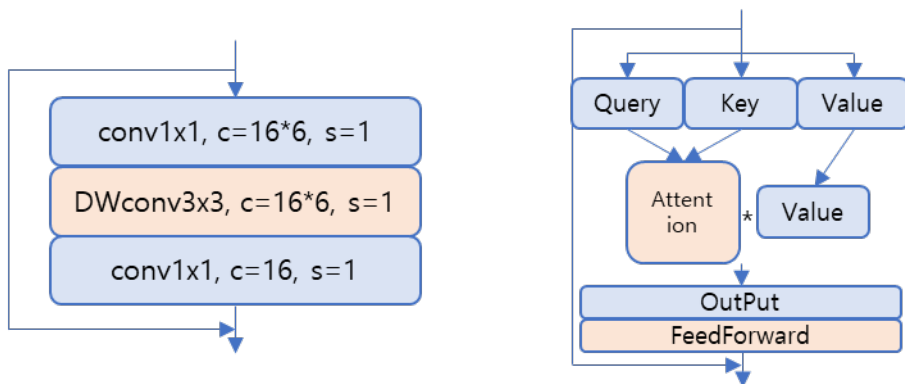




# 연구 문제 설정

## Naïve Approach

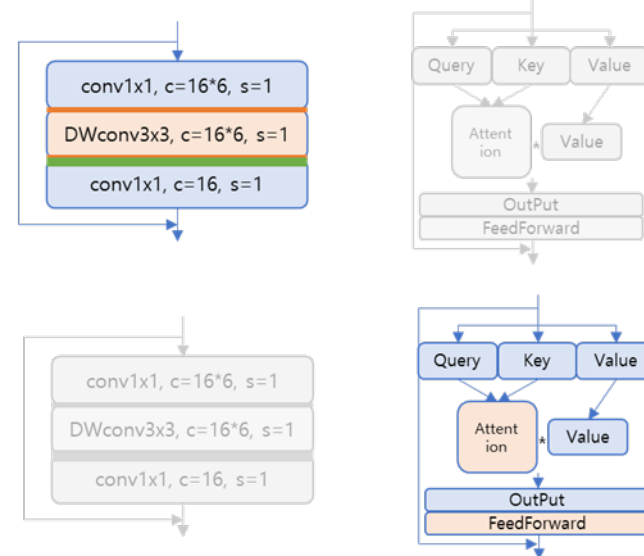
### Approach 1. Convex Combination



$$\alpha * y_{conv}(x) + (1 - \alpha) * y_{att}(x)$$

where  $\alpha \in (0, 1)$ , trainable

### Approach 2. Random Sampling

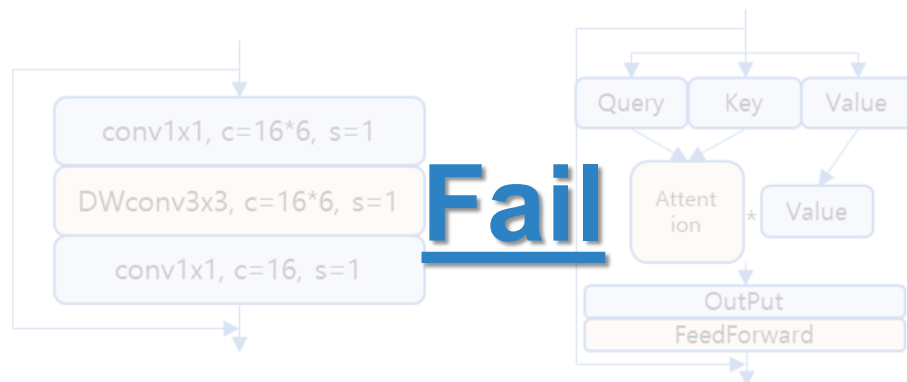


$\alpha \sim \text{UniformDistribution}$

# 연구 문제 설정

## Naïve Approach

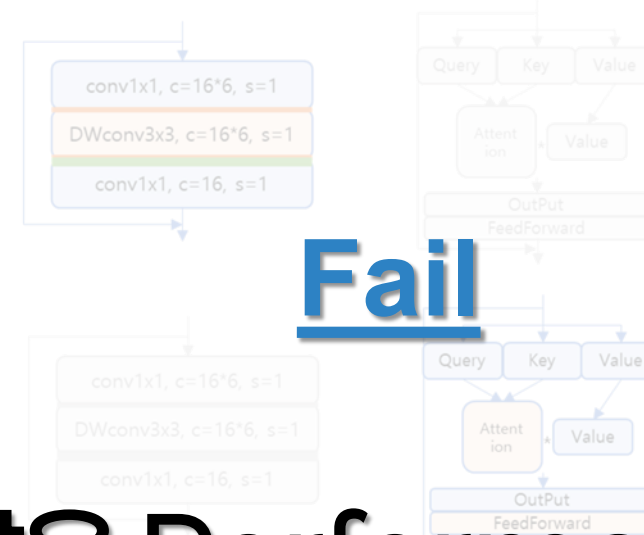
### Approach 1. Convex Combination



$\alpha * y_{att} + (1 - \alpha) * y_{ff}$   
 where  $\alpha \in (0, 1), trainable$

**$\alpha$ 의 수렴 실패**

### Approach 2. Random Sampling

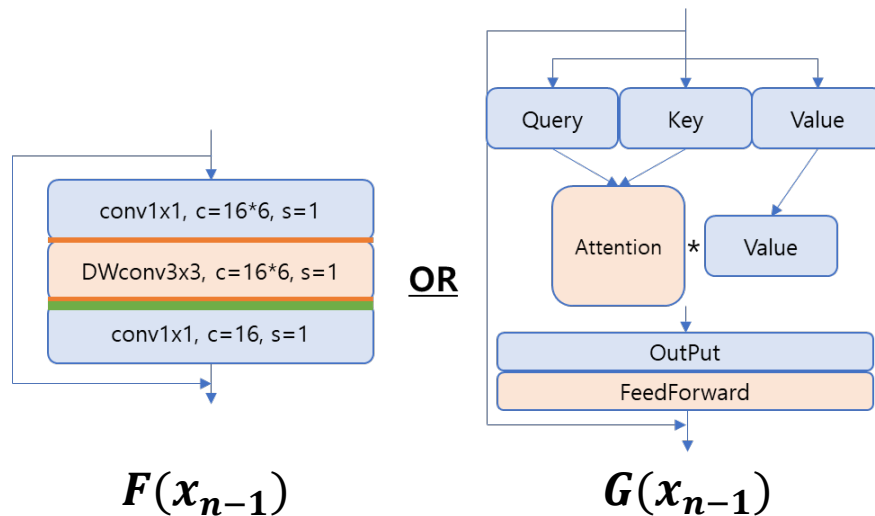


**낮은 Performance**  
 $\alpha \sim UniformDistribution$



# 신규 NAS 기술

## 매 스텝마다 연산자 가중 샘플링 (Weighted Samling)



## 샘플링 확률 실시간 업데이트

$$x_n = \begin{cases} (\alpha + T) \cdot F(x_{n-1}) & \text{if } s=1 \\ T: 1 - \alpha.detach() \\ (\beta + T) \cdot G(x_{n-1}) & \text{if } s=0 \\ T: 1 - \beta.detach() \end{cases}$$

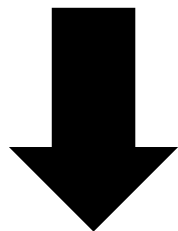
$s \sim \text{Bernoulli}(\alpha)$

$$\alpha + \beta = 1, \quad \alpha, \beta \in (0,1)$$

$\alpha$ : Convolution 샘플링 확률  
 $\beta$ : Transformer 샘플링 확률

# 신규 NAS 기술

$$\alpha \cdot F(x_{n-1}) + (1 - \alpha) \cdot G(x_{n-1})$$



$$(\alpha + T) \cdot F(x_{n-1})$$

$$T: 1 - \alpha.\text{detach}()$$

forward:

$$\begin{aligned} x_n &= (\alpha + (1 - \alpha)) * \text{ops}(x_{n-1}) \\ &= \text{ops}(x_{n-1}) \end{aligned}$$

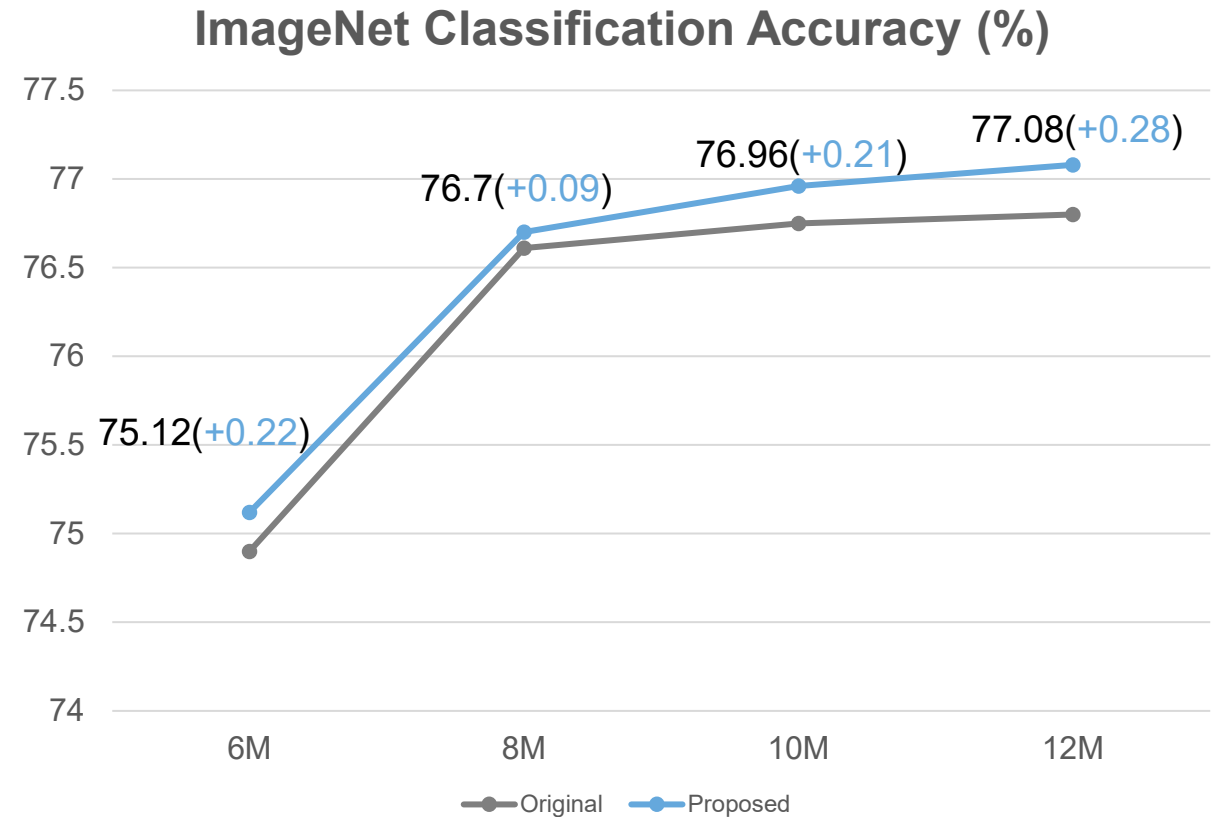
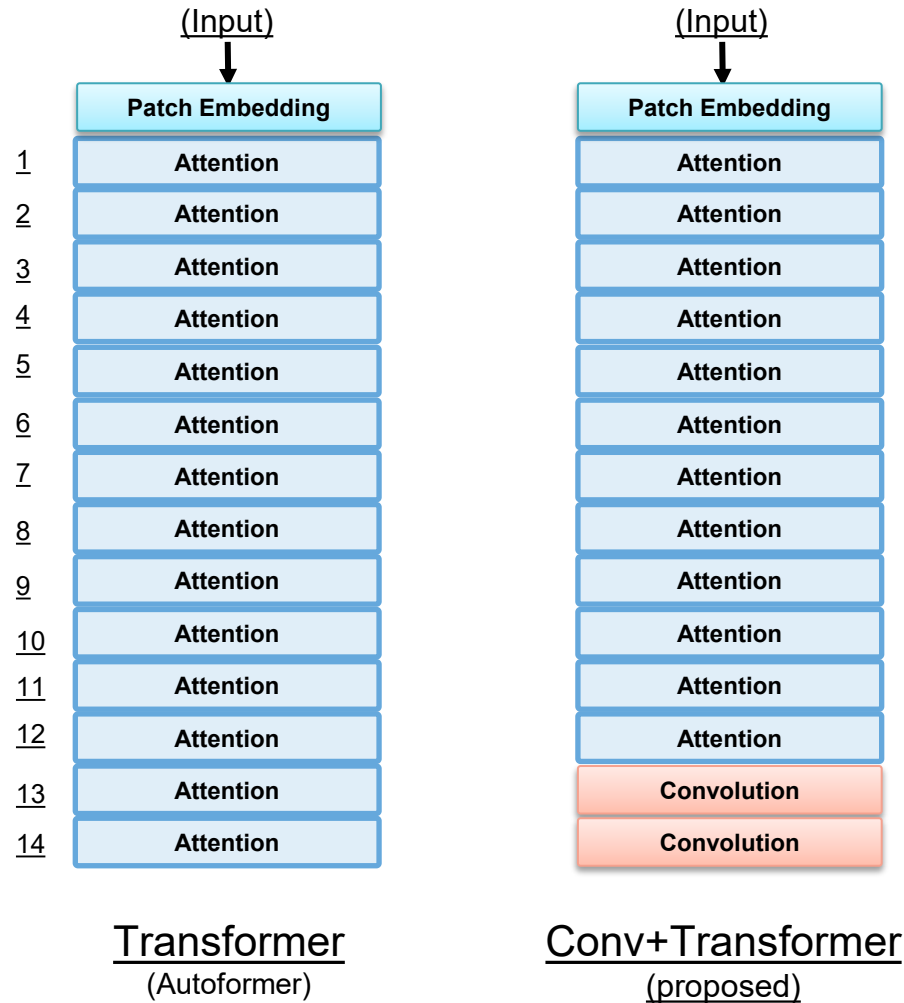
backward:

$$\begin{aligned} \frac{\partial x_n}{\partial \alpha} &= \text{ops}(x_{n-1}) + 0 - 0 \\ &= \text{ops}(x_{n-1}) \end{aligned}$$



# 신규 NAS 기술

## 실험 결과





# 감사합니다.

주관

ETRI  
한국전자통신연구원

( TANGO )

주최



과학기술정보통신부

IITP

정보통신기획평가원

후원

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(사)한국인공지능협회  
Korea Artificial Intelligence Association

SNUH  
서울대학교병원



고려대학교  
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홍익대학교  
HOONIK UNIVERSITY

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