#### CodeRL: Mastering Code Generation through Pretrained Models and Deep Reinforcement Learning

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Code and Model: https://github.com/salesforce/CodeRL

Paper: <u>https://arxiv.org/abs/2207.01780</u>

Blog: https://blog.salesforceairesearch.com/coderl/





1. CodeRL improves pretrained LMs for program synthesis by incorporating unit test signals in model training



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### 3. We extend CodeT5 with larger pretraining data, better pretraining objective, and larger model size



Pretraining Data

GitHub Code dataset



0,

Next token prediction



Up to 770M

# We evaluate CodeRL on two benchmarks: one with competitive programming tasks and another with beginner-level programming tasks



#### Solution Program



Write a python function to check if a given number is one less than twice its reverse. Your code should satisfy these tests:

assert check(70) == False assert check(23) == False assert check(73) == True

> Write a Python function to sort the given array by using merge sort. Your code should satisfy these tests:

assert merge\_sort([3, 4, 2, 6, 5, 7, 1, 9]) == [1, 2, 3, 4, 5, 6, 7, 9] assert merge\_sort([7, 25, 45, 78, 11, 33, 19]) == [7, 11, 19, 25, 33, 45, 78] assert merge\_sort([3, 1, 4, 9, 8]) == [1, 3, 4, 8, 9]

APPS

MBPP

#### CodeRL achieves new SoTA performance in programming tasks in both APPS and MBPP (zero-shot)

#### Performance by pass@k on APPS



Model	Size	pass@80
GPT	224M	7.2
GPT	422M	12.6
GPT	1B	22.4
GPT	4B	33.0
GPT	8B	40.6
GPT	68B	53.6
GPT	137B	61.4
CodeRL+CodeT5 (ZS)	770M	63.0

APPS

**MBPP** 

### CodeRL models show improvement over baseline models by generating more functionally correct programs



CodeRL models show improvement over baseline models by generating more functional correct programs Allen has a LOT of money. He has \$n\$ dollars in the bank. For security reasons, he wants to withdraw it in cash (we will not disclose the reasons here). The denominations for dollar bills are \$1\$, \$5\$, \$10\$, \$20\$, \$100\$. What is the minimum number of bills Allen could receive after withdrawing his entire balance? CodeRL models show improvement over baseline models by generating more functional correct programs

Example Unit Tests				
Input:	125			
Output:	3			
Input:	43			
Output:	5			
• • •				

**Ground-truth Program** n = int(input()) ans = 0ans += n/(100)n %= 100 ans += (n // 20) n %= 20 ans += n //10 n %= 10 9 ans += n//510 11  $n_{8} = 5$ 12 ans += n//113 14 print(ans) **Example tests: Passed** Hidden tests: Passed

	Hidden Unit Tests
Input: Output:	74 8
Input: Output: 	82655 830

CodeT5 Output Program		
1	<pre>n = int(input())</pre>	
2	<pre>a = defaultdict(int)</pre>	
4	for i in manage(1, 101).	
6	a[i] = n // i	
7	n = n % i	
8	ans = 0	
10	for i in range(1, 101):	
11	if n < a[i]:	
13	n -= a[i]	
14	print(ang)	
Example tests: Failed		

Hidden tests: Failed

CodeRL models show improvement over baseline models by generating more functional correct programs



In summary, CodeRL is a general framework that integrates pretrained LMs and RL holistically for program synthesis





**RL framework** 

Pretrained LMs such as CodeT5

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#### THANK YOU!

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