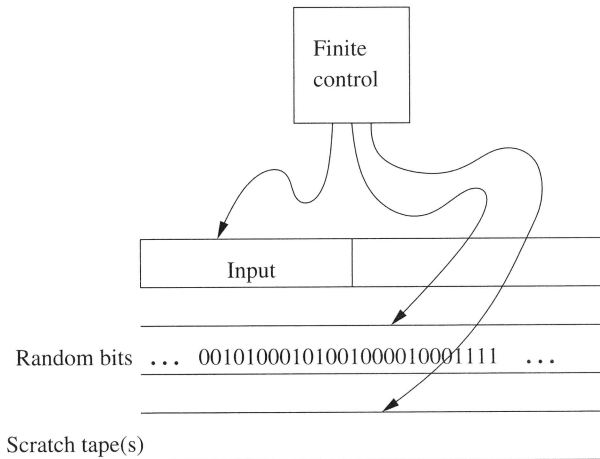


Language Classes Based on Randomization

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CIS 890

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Randomized Turing Machines



Quicksort

A RTM M does as follows on input w where $|w| = m$:

1. Use about $O(\log m)$ new random bits on Tape 2 to pick a random number n between $1 \dots m$. The n^{th} symbol in w is the pivot p .
2. Put p on Tape 3.
3. Scan w , copying each symbol x to Tape 4 if $x \leq p$.
4. Scan w , copying each symbol y to Tape 5 if $y > p$.
5. Overwrite w on Tape 1 with contents of Tape 4 and then 5, placing a marker between them.
6. If either or both sublists have more than one element, recursively sort them by the same algorithm.

Running time is in $O(n \log n)$

Language of Randomized Turing Machines

- ▶ Each “branch” of a RTM has a probability
- ▶ On a given input w , a RTM M :
 - ▶ may have different runtime behavior
 - ▶ may not halt
- ▶ Each input w to M has some probability of acceptance
- ▶ Time and space complexity can be measured using the worst case computation branch

Example Probability Calculation

	00	01	10	11	B0	B1
$\rightarrow q_0$	$q_1 00RS$	$q_3 01SR$	$q_2 10RS$	$q_3 11SR$		
q_1	$q_1 00RS$				$q_4 B0SS$	
q_2			$q_2 10RS$		$q_4 B0SS$	
q_3	$q_3 00RR$			$q_3 11RR$	$q_4 B0SS$	$q_4 B1SS$
$*q_4$						

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q_0

0	0	0	0	B
0	1	1	0	1

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	0	0	0	0	B
q_4	0	1	1	0	1

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	00	01	10	11	B0	B1
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q_2

1	1	1	1	B
0	1	1	0	1

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	00	01	10	11	B0	B1
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q_0

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Homogeneous Input w (0^i or 1^i)

$$\frac{1}{2} + \frac{1}{2}2^{-i} = \frac{1}{2} + 2^{-(i+1)}$$

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Heterogeneous Input w (both 0's and 1's)

- ▶ if first random bit is 0 then w is never accepted
- ▶ Otherwise the probability of acceptance is $2^{-|w|}$

$$\frac{1}{2} 2^{-|w|} = 2^{-(|w|+1)}$$

Example Probability Calculation

	00	01	10	11	$B0$	$B1$
$\rightarrow q_0$	$q_1 00RS$	$q_3 01SR$	$q_2 10RS$	$q_3 11SR$		
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Conclusion: We can compute a probability of acceptance of any given string by any given RTM.

\mathcal{RP} Random Polynomial

A language L is said to be in \mathcal{RP} if it is accepted by a RTM M such that on input w :

1. If $w \notin L$, then the probability that M accepts w is 0.
2. If $w \in L$, then the probability that M accepts w is at least $1/2$.
3. There exists a polynomial $T(n)$ where $n = |w|$ such that all runs of M halt after at most $T(n)$ steps.

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Polynomial-time Monte-Carlo TMs

The class of languages for which membership can be determined in polynomial time by a RTM with no false acceptances and less than half of the rejections are false rejections

Recognizing Languages in \mathcal{RP}

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Recognizing Languages in \mathcal{RP}

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- ▶ Repeating the tests will take polynomial time as c is a constant and one run of a RTM takes polynomial time

Theorem 11.16: If L is in \mathcal{RP} , then for any constant $c > 0$, no matter how small, there is a polynomial-time randomized algorithm that renders a decision whether its given input w is in L , makes no false-positive errors, and makes false-negative errors with probability no greater than c .

ZPP Zero-Error, Probabilistic, Polynomial

- ▶ ZPP is based on a RTM that always halts, and has an expected time to halt that is some polynomial in the length of the input.

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- ▶ ZPP is based on a RTM that always halts, and has an expected time to halt that is some polynomial in the length of the input.
- ▶ Similar to the definition of P except ZPP allows the TM to involve randomness, and the expected running time rather than worst-case is measured
- ▶ **Defn:** *Las-Vegas Turning Machines* are TMs that always give the correct answer, but whose running time varies depending on the values of some random bits.

ZPP Zero-Error, Probabilistic, Polynomial

- ▶ ZPP is based on a RTM that always halts, and has an expected time to halt that is some polynomial in the length of the input.
- ▶ Similar to the definition of \mathcal{P} except ZPP allows the TM to involve randomness, and the expected running time rather than worst-case is measured
- ▶ **Defn:** *Las-Vegas Turning Machines* are TMs that always give the correct answer, but whose running time varies depending on the values of some random bits.
- ▶ A language $L \in ZPP$ if it's accepted by a *Las-Vegas Turing Machine* with a polynomial expected running time.

Relationship between \mathcal{RP} and \mathcal{ZPP}

If $L \in \mathcal{ZPP}$, then so is \bar{L}

- ▶ Assume M is a polynomial-expected-time Las-Vegas TM such that $L \in L(M)$
- ▶ $\bar{L} \in L(M')$ such that all accepting states in M are changed to halting without acceptance states in M' and all non-accepting halting states in M are changed to accepting and halt states in M' .

Relationship between \mathcal{RP} and \mathcal{ZPP}

Theorem 11.17: $\mathcal{ZPP} = \mathcal{RP} \cap \text{co-}\mathcal{RP}$

Proof: $\mathcal{RP} \cap \text{co-}\mathcal{RP} \subseteq \mathcal{ZPP}$

Suppose $L \in \mathcal{RP} \cap \text{co-}\mathcal{RP}$. Therefore L and \bar{L} have Monte-Carlo TM's S and T , each with a polynomial running time. Assume $p(n)$ bounds the running time of both machines. We design a Las-Vegas TM M for L as follows:

1. Run S on the input; if it accepts, then M accepts and halts.
2. If not, run T on the input. If it accepts, then M halts without accepting. Otherwise, M returns to step (1)

Relationship between \mathcal{RP} and \mathcal{ZPP}

- ▶ Clearly M accepts w if $w \in L$ and rejects w only if $w \notin L$.
- ▶ The expected running time of round one is $2p(n)$. Step (1) has a 50% chance of leading to acceptance and Step (2) has a 50% chance of leading to rejection so the expected running time of M is no more than

$$2p(n) + \frac{1}{2}2p(n) + \frac{1}{4}2p(n) + \frac{1}{8}2p(n) + \cdots = 4p(n)$$

Relationship to classes \mathcal{P} and \mathcal{NP}

Proof: Assume $L \in \mathcal{ZPP}$ and show $L \in \mathcal{RP}$ and $L \in \text{co-}\mathcal{RP}$.

We construct a Monte-Carlo TM M_2 as follows:

- ▶ M_2 simulates S for $2p(n)$ steps.
- ▶ If S accepts then so does M_2 ; otherwise M_2 rejects.

Observations:

- ▶ Suppose $w \notin L$. Then S will not accept w nor will M_2 .
- ▶ Suppose $w \in L$. Then S will eventually accept w but not necessarily within $2p(n)$ steps.
- ▶ However the probability of S accepting w within $2p(n)$ steps is at least $1/2$ and therefore the probability of M_2 accepting w is at least $1/2$.
- ▶ Thus M_2 is polynomial-time-bounded Monte-Carlo TM, so $L \in \mathcal{RP}$. □

Relationship to classes \mathcal{P} and \mathcal{NP}

Theorem 11.18: $\mathcal{P} \subseteq \mathcal{ZPP}$

Proof: Any deterministic, polynomial-time bounded TM is also a Las-Vegas, polynomial-time bounded TM, that happens not to use its ability to make random choices \square

Relationship to classes \mathcal{P} and \mathcal{NP}

Theorem 11.19: $\mathcal{RP} \subseteq \mathcal{NP}$

Proof: Suppose we have a polynomial-time-bounded Monte-Carlo TM M_1 for a language L . When M_1 examines a random bit, M_2 non-deterministically chooses both possible values for the bit and writes it to its own tape which simulates the random tape of M_1 . M_2 accepts whenever M_1 accepts and does not accept otherwise.

Suppose $w \in L$. Since the chance M_1 accepts w is 50%, there must exist a sequence of random bits that leads to acceptance of w . M_2 will eventually choose this sequence and therefore also accepts when M_1 does. Thus $w \in L(M_2)$.

However, if $w \notin L$ then there are no sequence of random bits that lead to acceptance in M_1 and therefore no sequence of choices of M_2 will lead to acceptance. Therefore $w \notin L(M_2)$.

Relationship to classes \mathcal{P} and \mathcal{NP}

