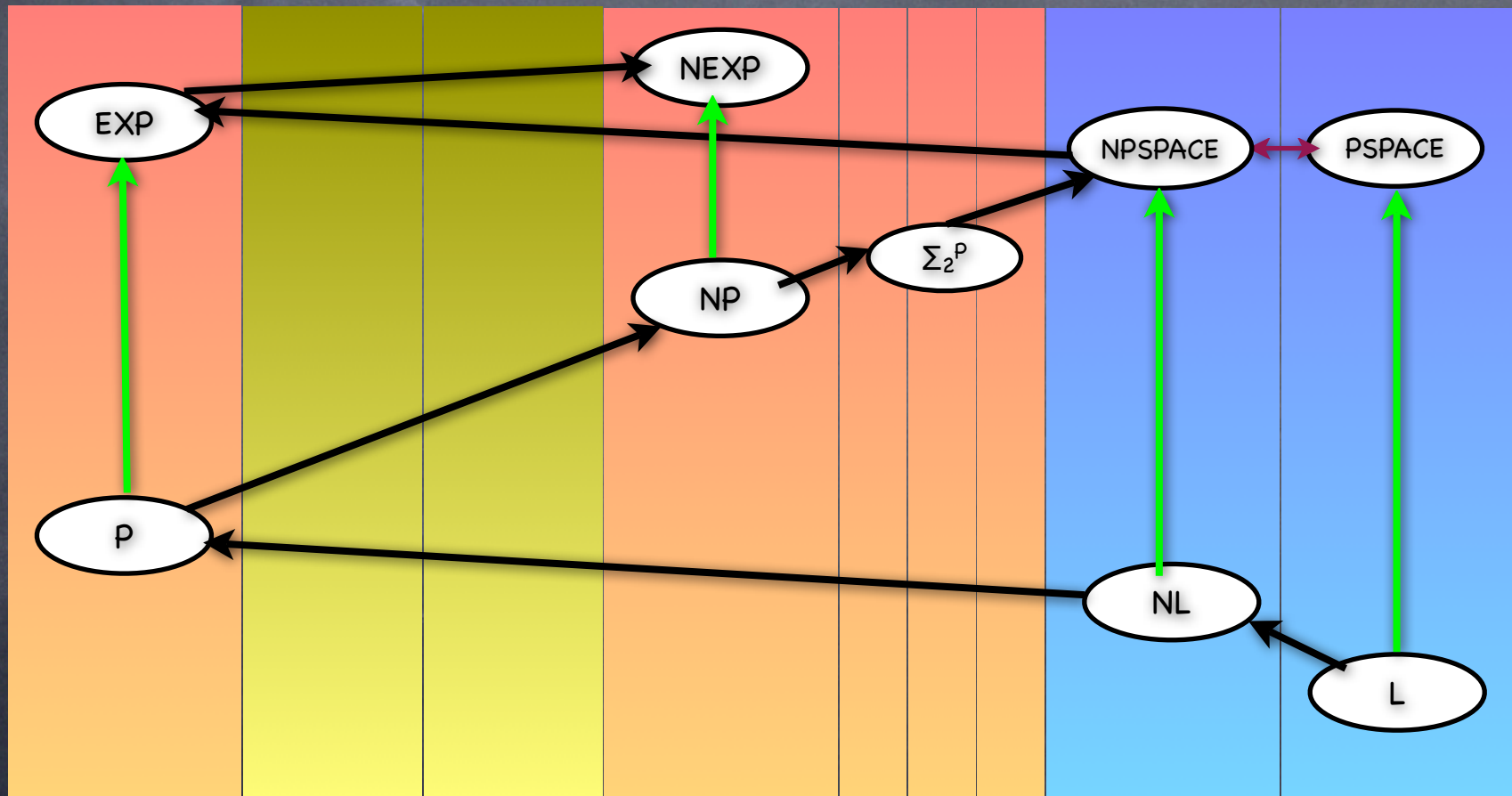


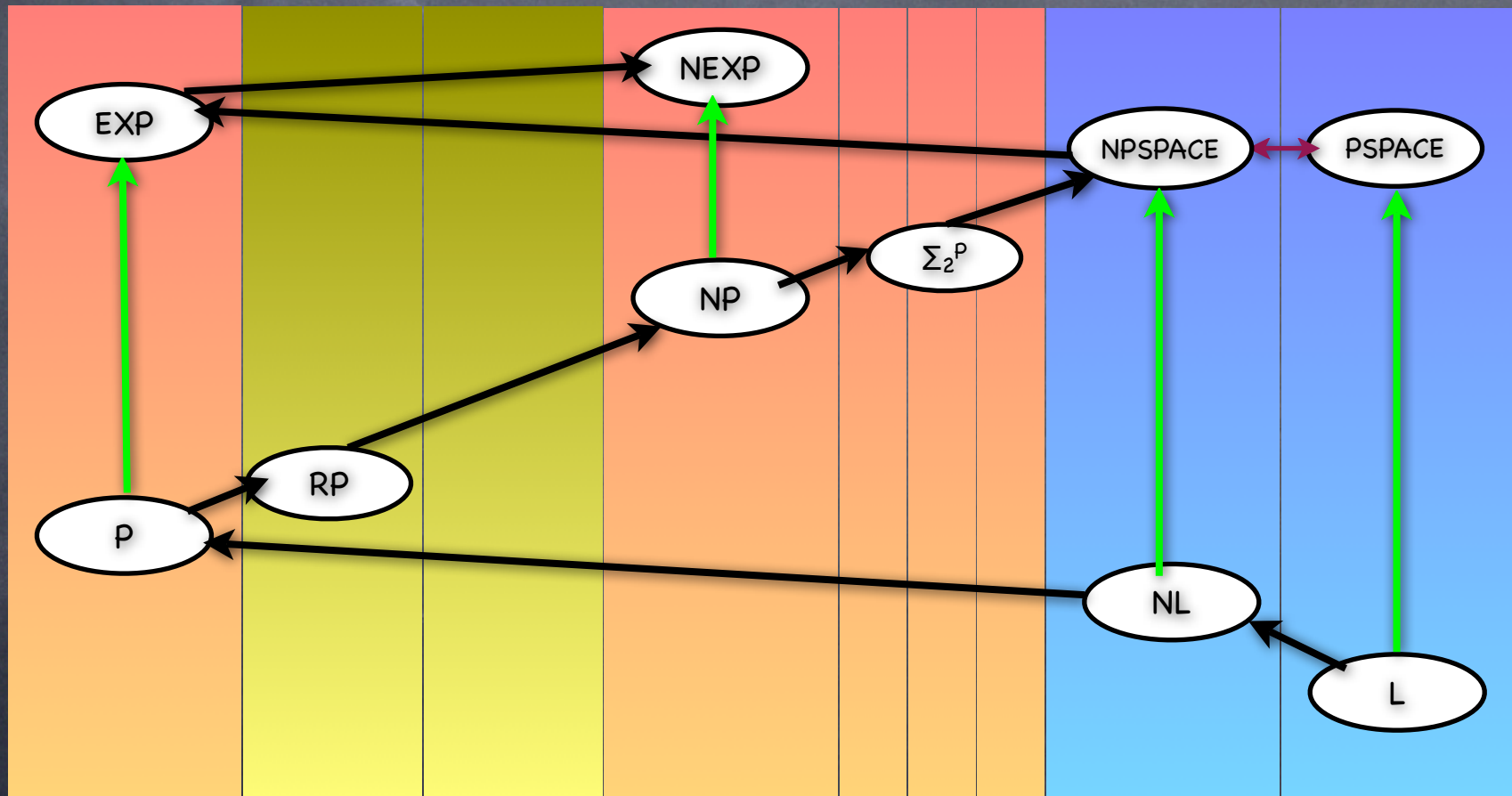
# Probabilistic Computation

Lecture 13  
BPP, ZPP

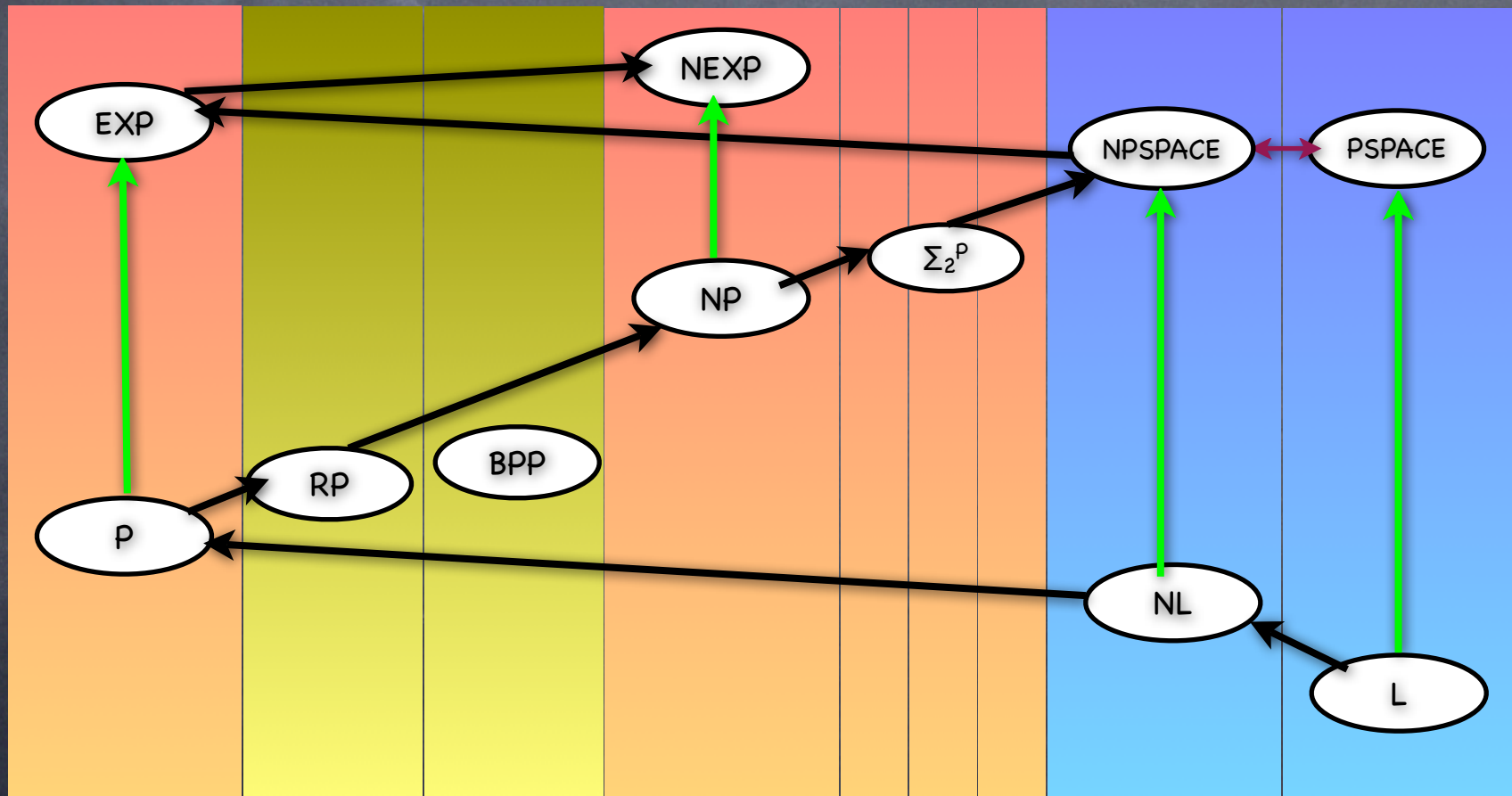
# Zoo



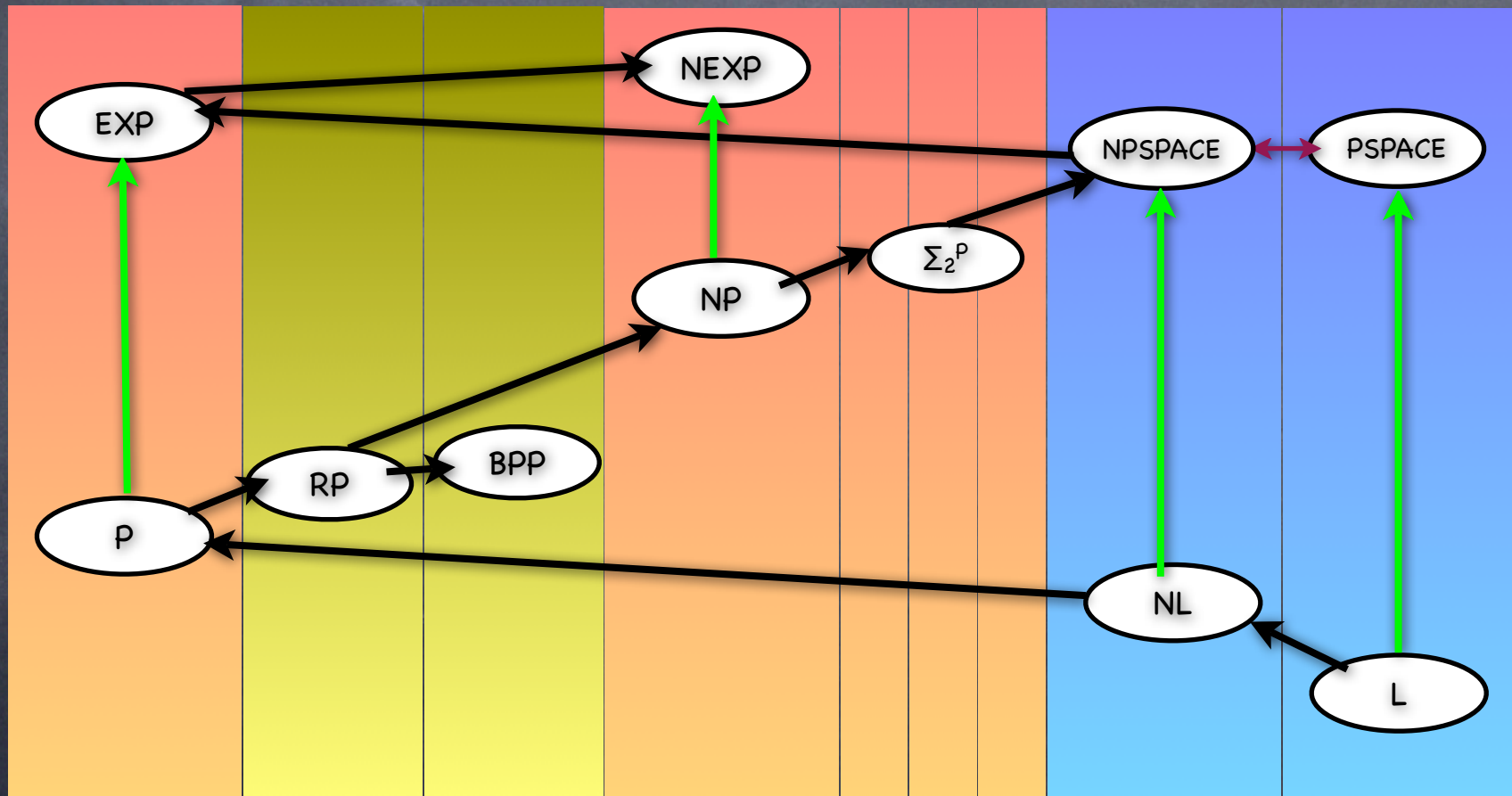
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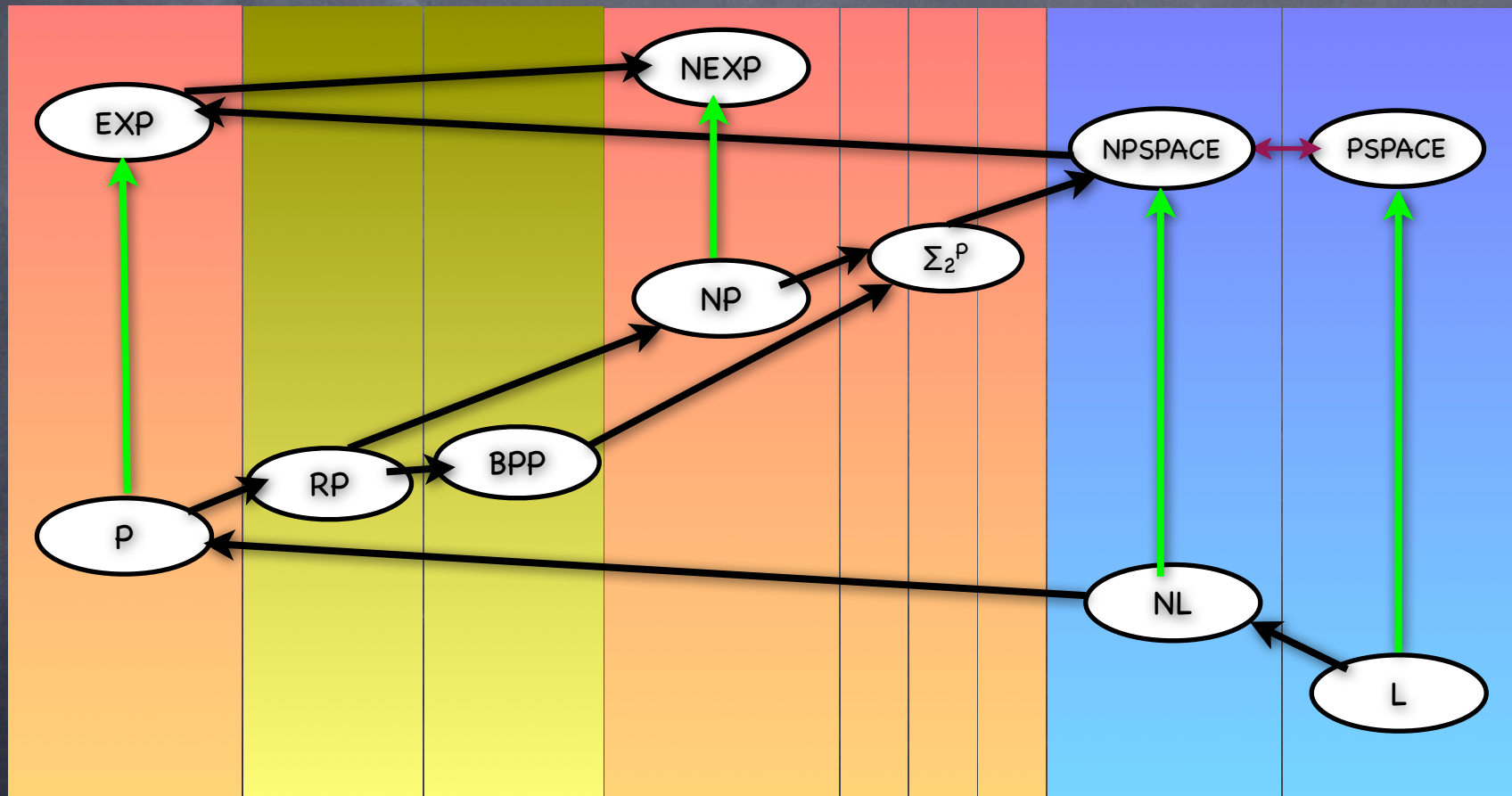
# Zoo



# Zoo



# Zoo



# Some Probabilistic Algorithmic Concepts

# Some Probabilistic Algorithmic Concepts

- Sampling to determine some probability
  - Checking if determinant of a symbolic matrix is zero: Substitute random values for the variables and evaluate
  - Polynomial Identity Testing: polynomial given as an arithmetic circuit. Like above, but values can be too large. So work over a random modulus.
- Random Walks (for sampling)
  - Monte Carlo algorithms for calculations
  - Reachability tests



# Random Walks

# Random Walks

- Which nodes does the walk touch and with what probability?
  - How do these probabilities vary with number of steps
- Analyzing a random walk
  - Probability Vector:  $\mathbf{p}$
  - Transition probability matrix:  $M$
  - One step of the walk:  $\mathbf{p}' = M\mathbf{p}$
  - After  $t$  steps:  $\mathbf{p}^{(t)} = M^t\mathbf{p}$

# Space-Bounded Probabilistic Computation

# Space-Bounded Probabilistic Computation

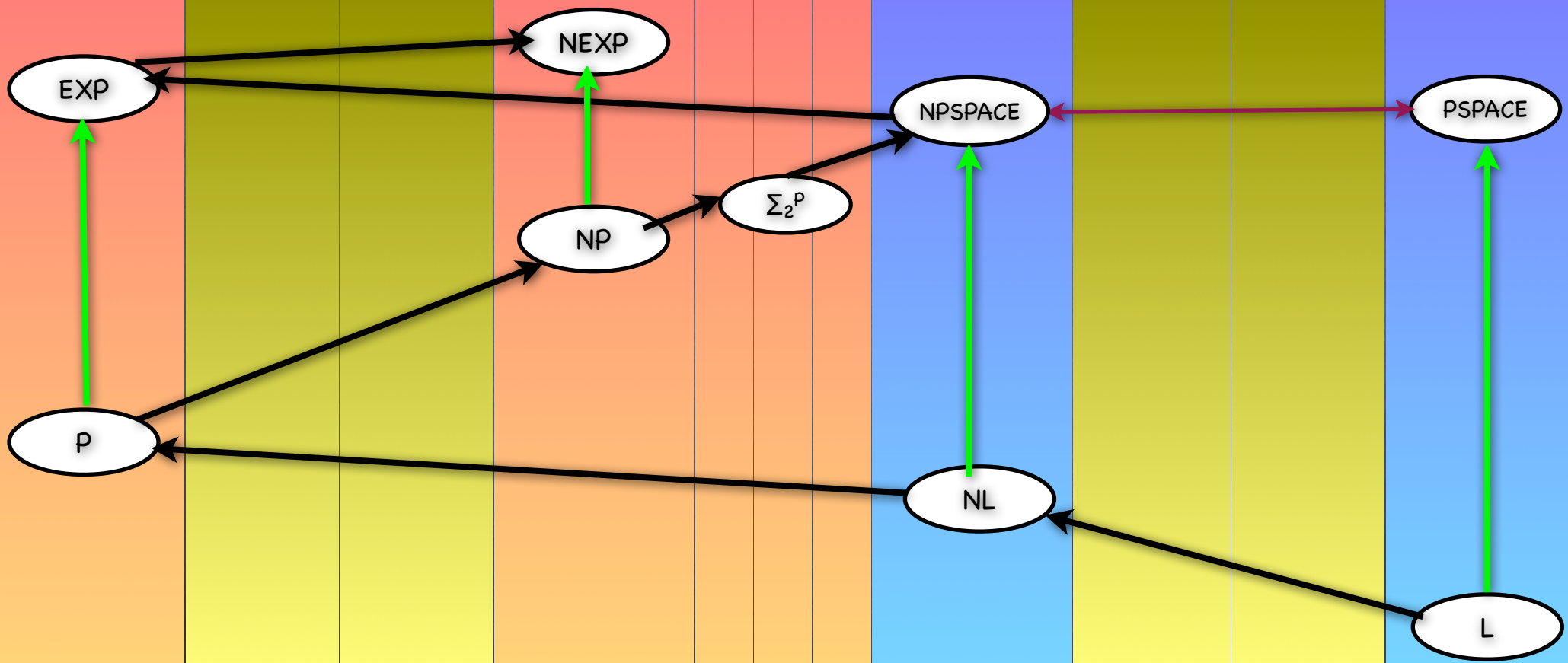
- PL, RL, BPL
  - Logspace analogues of PP, RP, BPP
- Note:  $RL \subseteq NL$ ,  $RL \subseteq BPL$ 
  - Recall  $NL \subseteq P$  (because  $PATH \in P$ )
  - So  $RL \subseteq P$
  - In fact  $BPL \subseteq P$

$\text{BPL} \subseteq \text{P}$

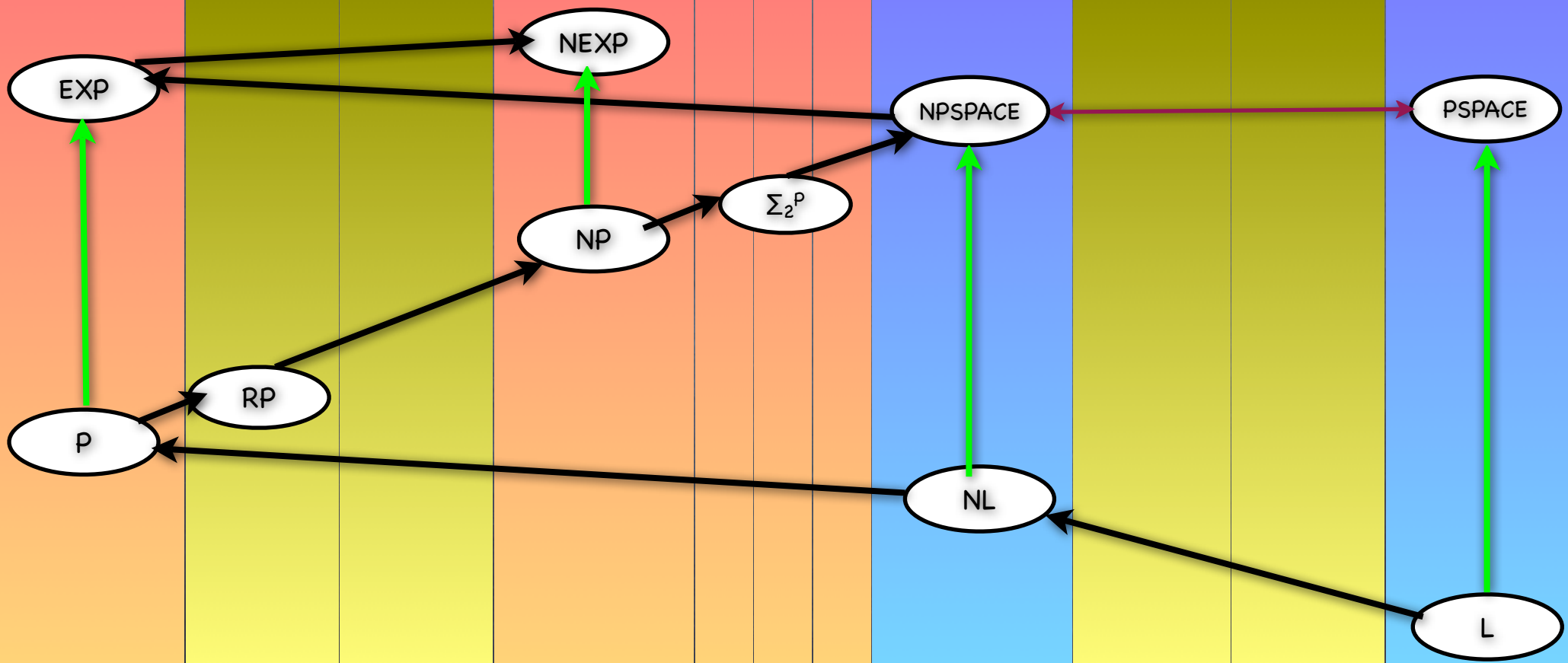
$$\text{BPL} \subseteq \text{P}$$

- Consider the BPL algorithm, on input  $x$ , as a random walk over states
  - Construct the transition matrix  $M$
  - Size of graph is  $\text{poly}(n)$ , probability values are 0, 0.5 and 1
  - Calculate  $M^t$  for  $t = \text{max running time} = \text{poly}(n)$
  - Accept if  $(M^t P)_{\text{accept}} > 2/3$

# Zoo

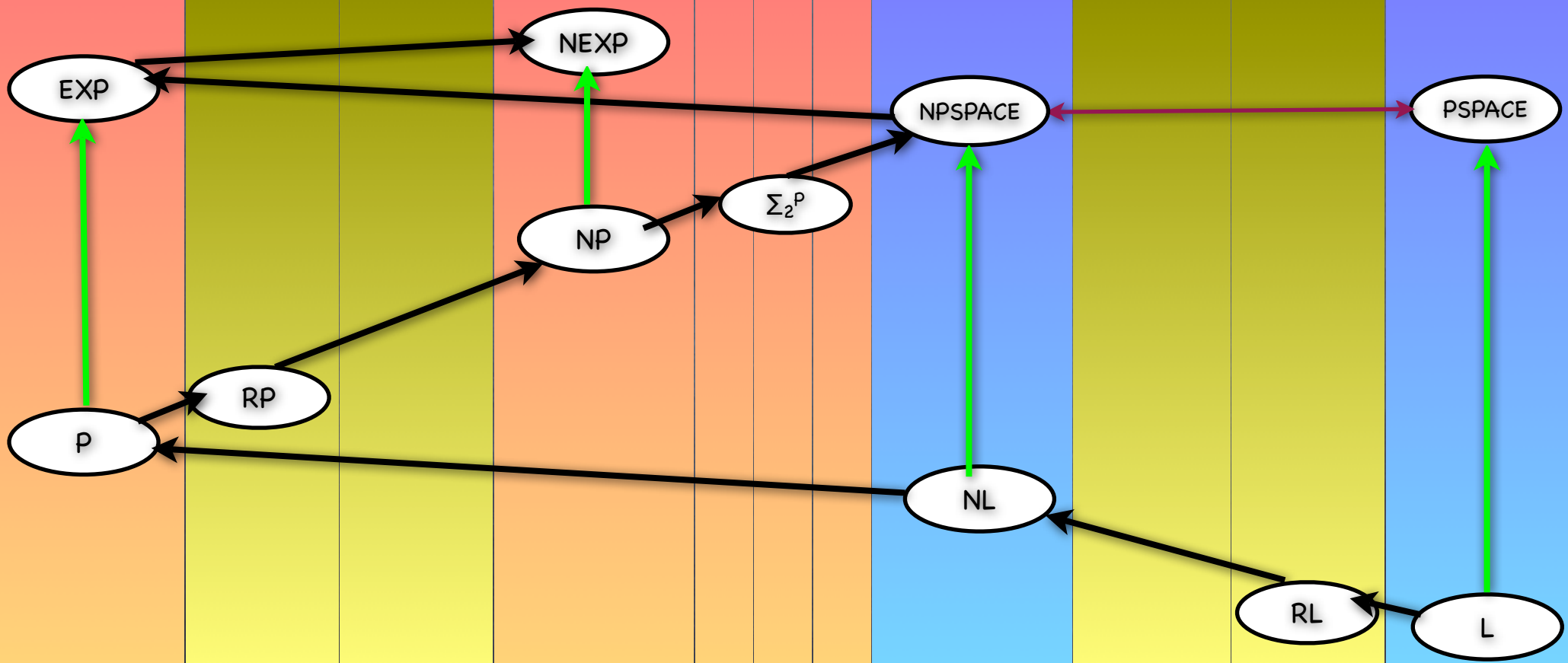


# Zoo

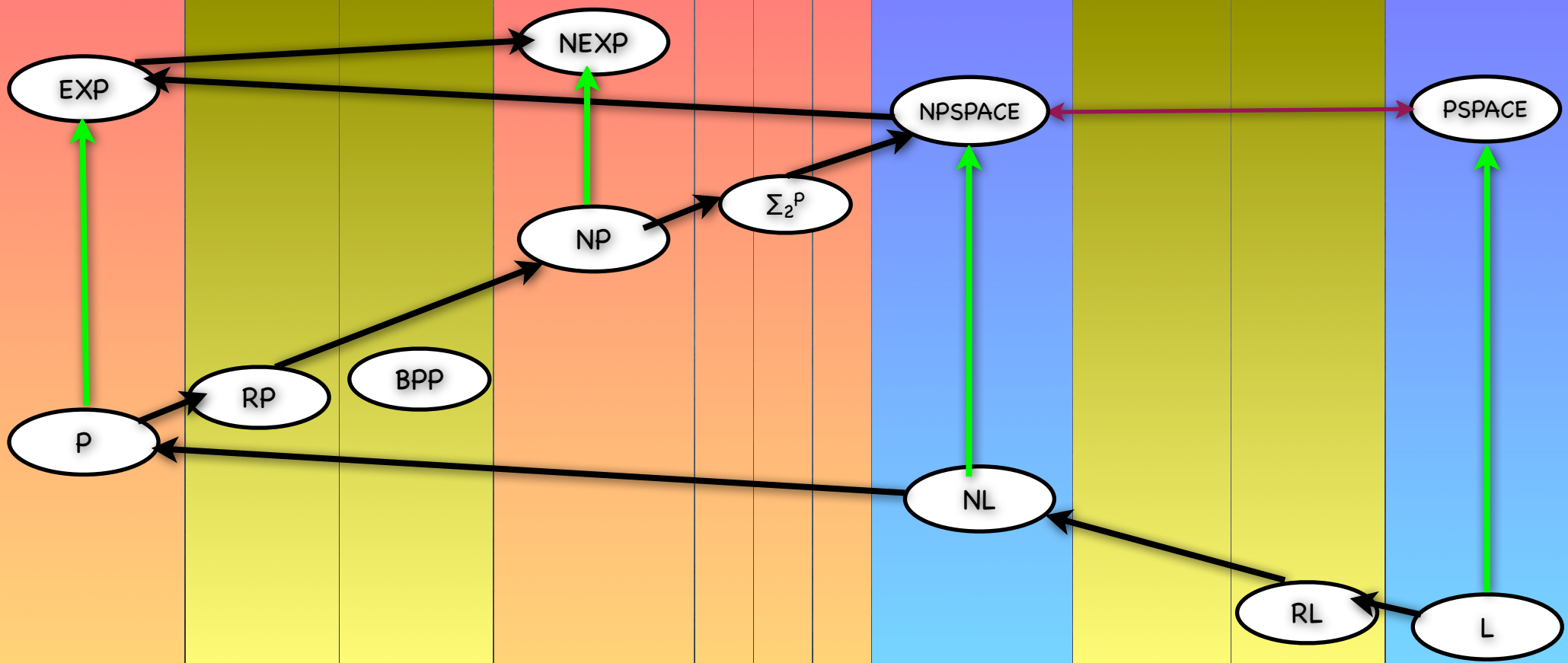




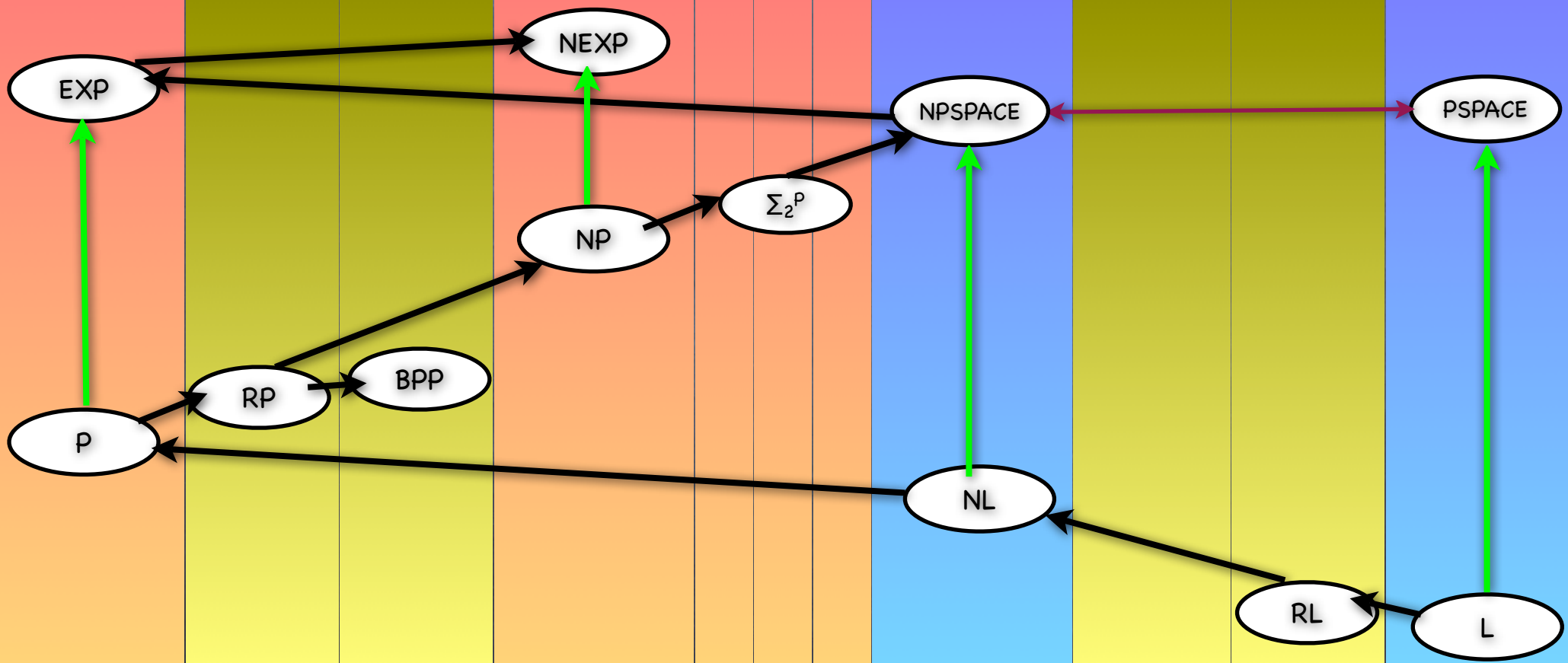
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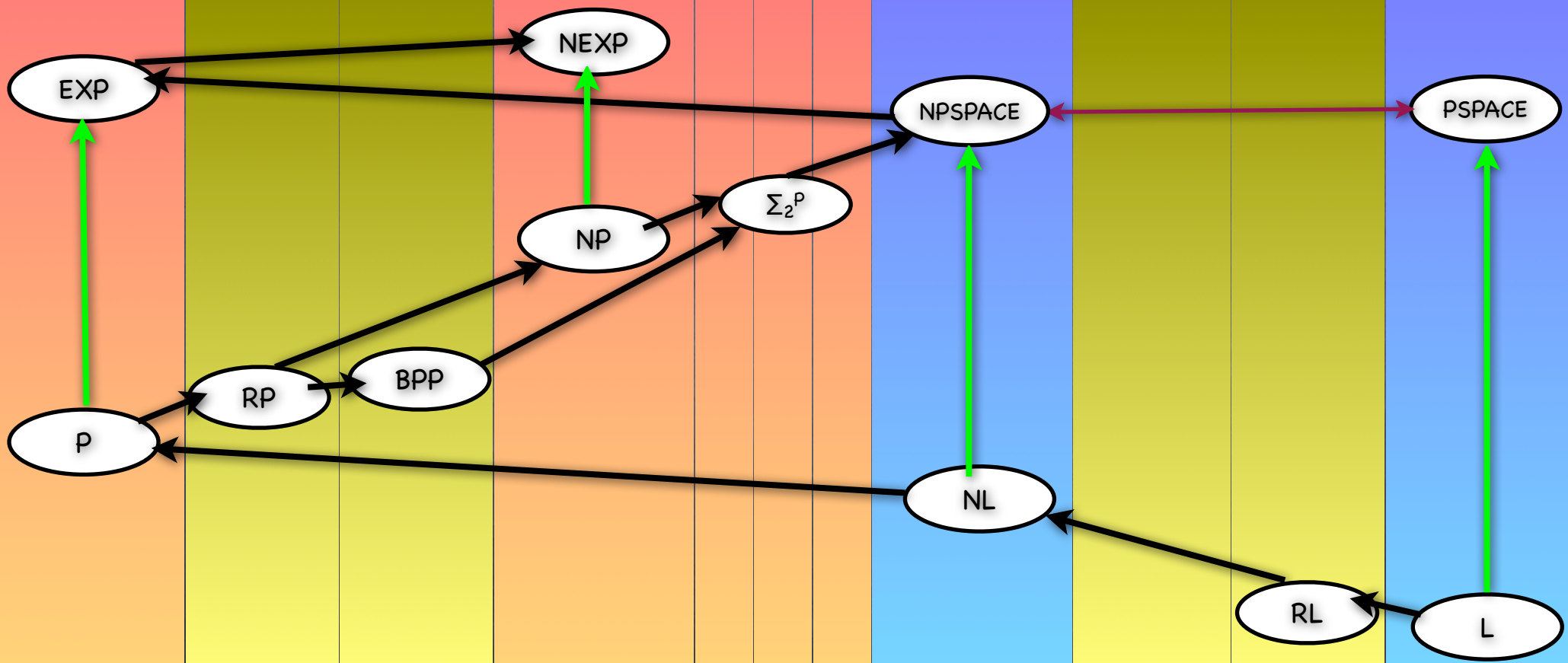
# Zoo



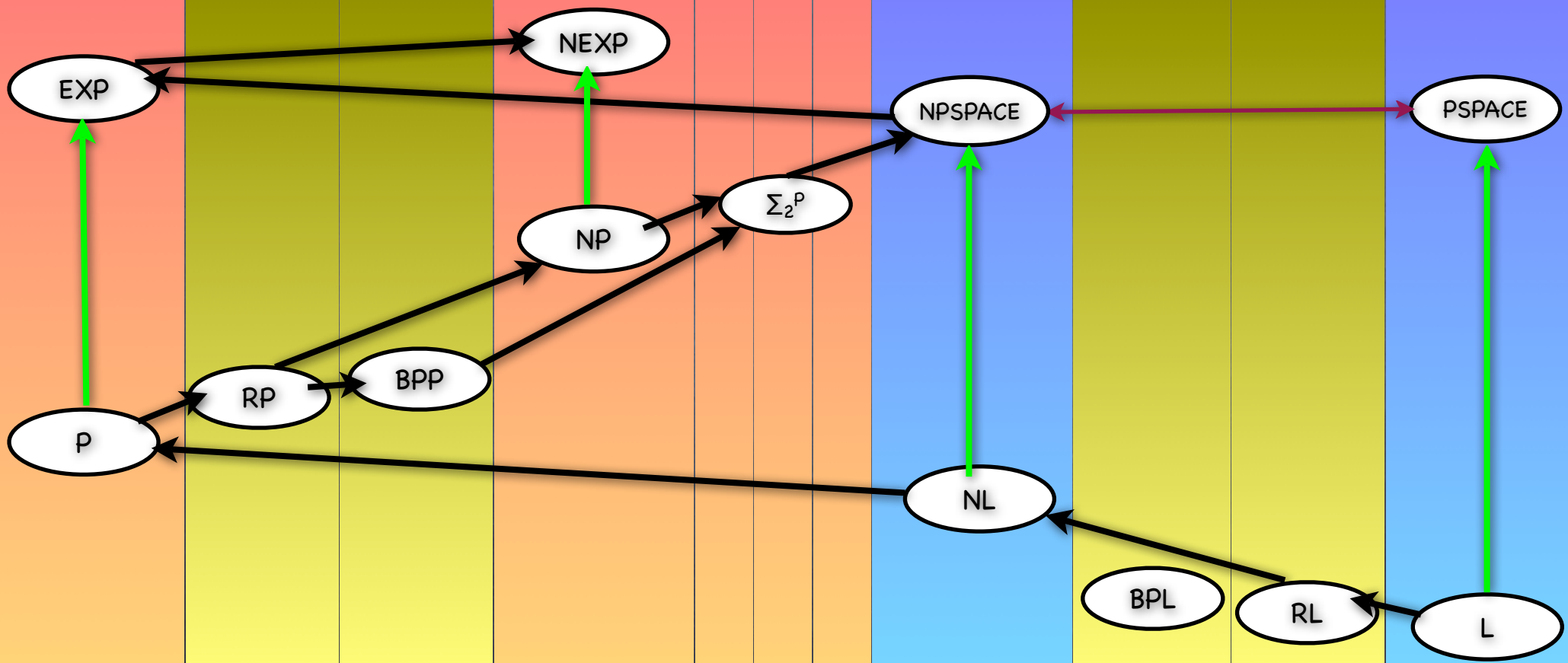
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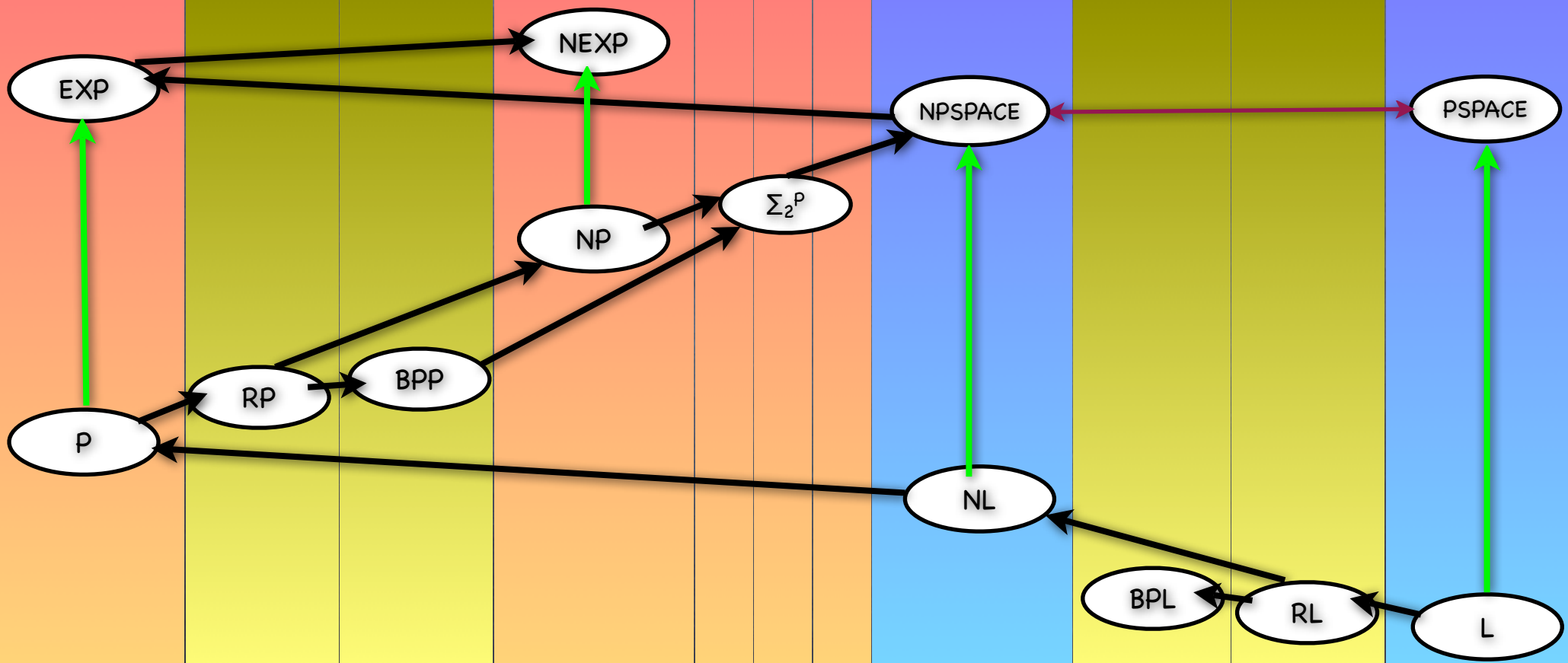
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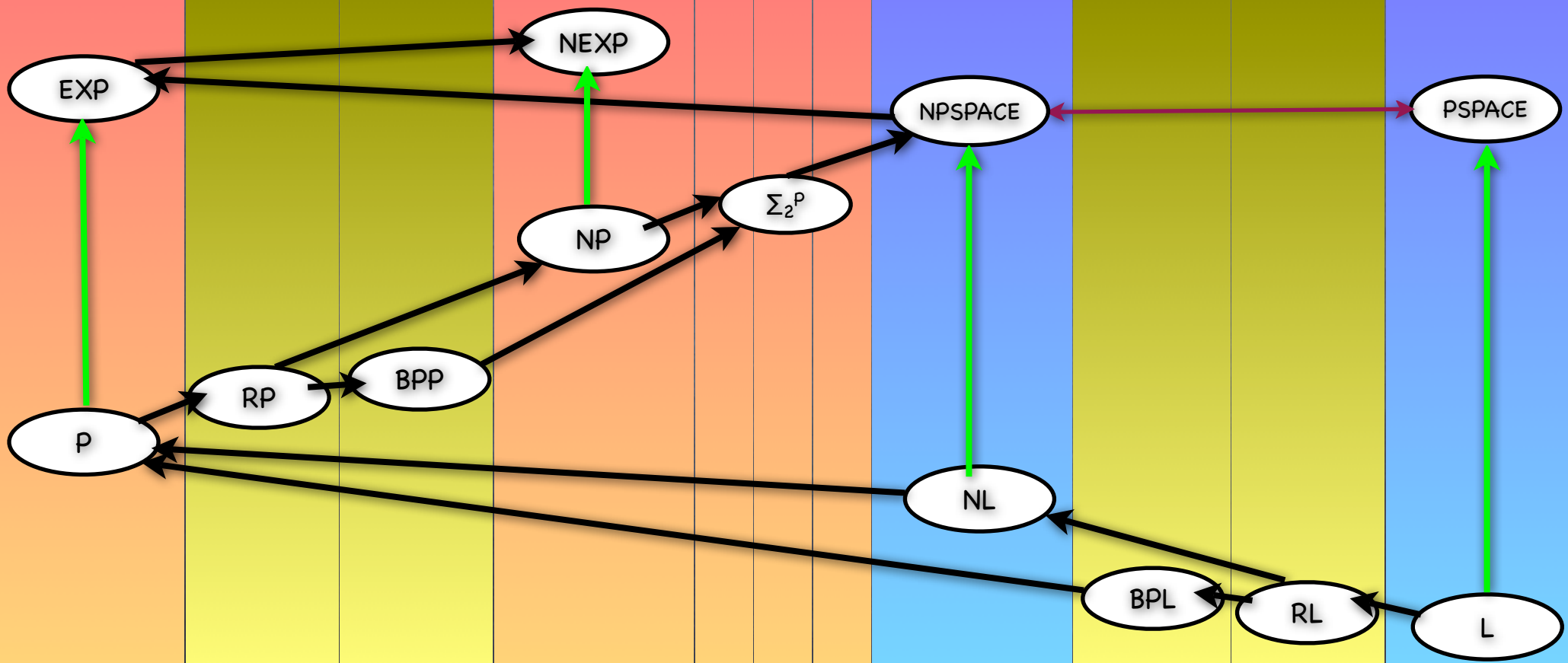
# Zoo



# Zoo



# Zoo



# Expected Running Time



# Expected Running Time

- Running time is a random variable too
  - As is the outcome of yes/no
- Ask for running time being polynomial only in expectation, or with high probability
- Las Vegas algorithms: only expected running time is polynomial; but when it terminates, it produces the correct answer
  - Zero error probability

# Zero-Error Computation

# Zero-Error Computation

- e.g. A simple algorithm for finding median in expected linear time
  - (There are non-trivial algorithms to do it in deterministic linear time. Simple sorting takes  $O(n \log n)$  time.)
- Procedure Find-element( $L, k$ ) to find  $k^{\text{th}}$  smallest element in list  $L$ 
  - Pick random element  $x$  in  $L$ . Scan  $L$ ; divide it into  $L_{>x}$  (elements  $> x$ ) and  $L_{<x}$  (elements  $< x$ ); also determine position  $m$  of  $x$  in  $L$ .
  - If  $m = k$ , return  $x$ . If  $m > k$ , call Find-element( $L_{<x}, k$ ), else call Find-element( $L_{>x}, k-m$ )
- Correctness obvious. Expected running time?

# Zero-Error Computation

# Zero-Error Computation

- Expected running time (worst case over all lists of size  $n$ , and all  $k$ ) be  $T(n)$
- Time for non-recursive operations is linear: say bounded by  $cn$ . Will show inductively  $T(n)$  at most  $4cn$  (base case  $n=1$ ).
- $T(n) \leq cn + 1/n [\sum_{n>j>k} T(j) + \sum_{0<j<k} T(n-j)]$
- $T(n) \leq cn + 1/n \cdot 4c [\sum_{j>k} j + \sum_{j<k} (n-j)]$  by inductive hypothesis
- $\sum_{j>k} j + \sum_{j<k} (n-j) = \sum_{j>k} j + (k-1)n - \sum_{j<k} j \leq \sum_j j + (k-1)n - 2 \sum_{j<k} j$ 
  - $\leq n^2/2 + (k-1)n - k(k-1) < n^2/2 + k(n-k) \leq 3/4 n^2$
  - $T(n) \leq cn + 3cn$  as required

# Zero-Error Computation

# Zero-Error Computation

- Las-Vegas Algorithms: Probabilistic algorithms with deterministic outcome (but probabilistic run time)
- $ZPTIME(T)$ : class of languages decided by a zero-error probabilistic TM, with expected running time at most  $T$
- $ZPP = ZPTIME(\text{poly})$ 
  - $ZPP = RP \cap \text{co-RP}$

$$\text{ZPP} \subseteq \text{RP}$$



# ZPP $\subseteq$ RP

- Truncate after “long enough,” and say “no”
- Do we still have bounded (one-sided) error?
- Will run for “too long” only with small probability
  - Because expected running time short
  - With high probability the running time does not exceed the expected running time by much
  - $\Pr[ x > a E[X] ] < 1/a$  (non-negative X)
  - **Markov's inequality**
- $\Pr[\text{error}]$  changes by at most  $1/a$  if truncated after  $a$  times expected running time

$$\text{RP} \cap \text{co-RP} \subseteq \text{ZPP}$$

$$\text{RP} \cap \text{co-RP} \subseteq \text{ZPP}$$

- If  $L \in \text{RP} \cap \text{co-RP}$  a ZPP algorithm for  $L$ :
  - Run both RP and coRP algorithms
  - If former says yes or latter says no, output that answer
  - Else, i.e., if former says no and latter yes, repeat
    - Expected number of repeats =  $O(1)$

Today

# Today

- Zoo
  - $BPL \subseteq P$
- Expected running time
- Zero-Error probabilistic computation
- $ZPP = RP \cap co-RP$