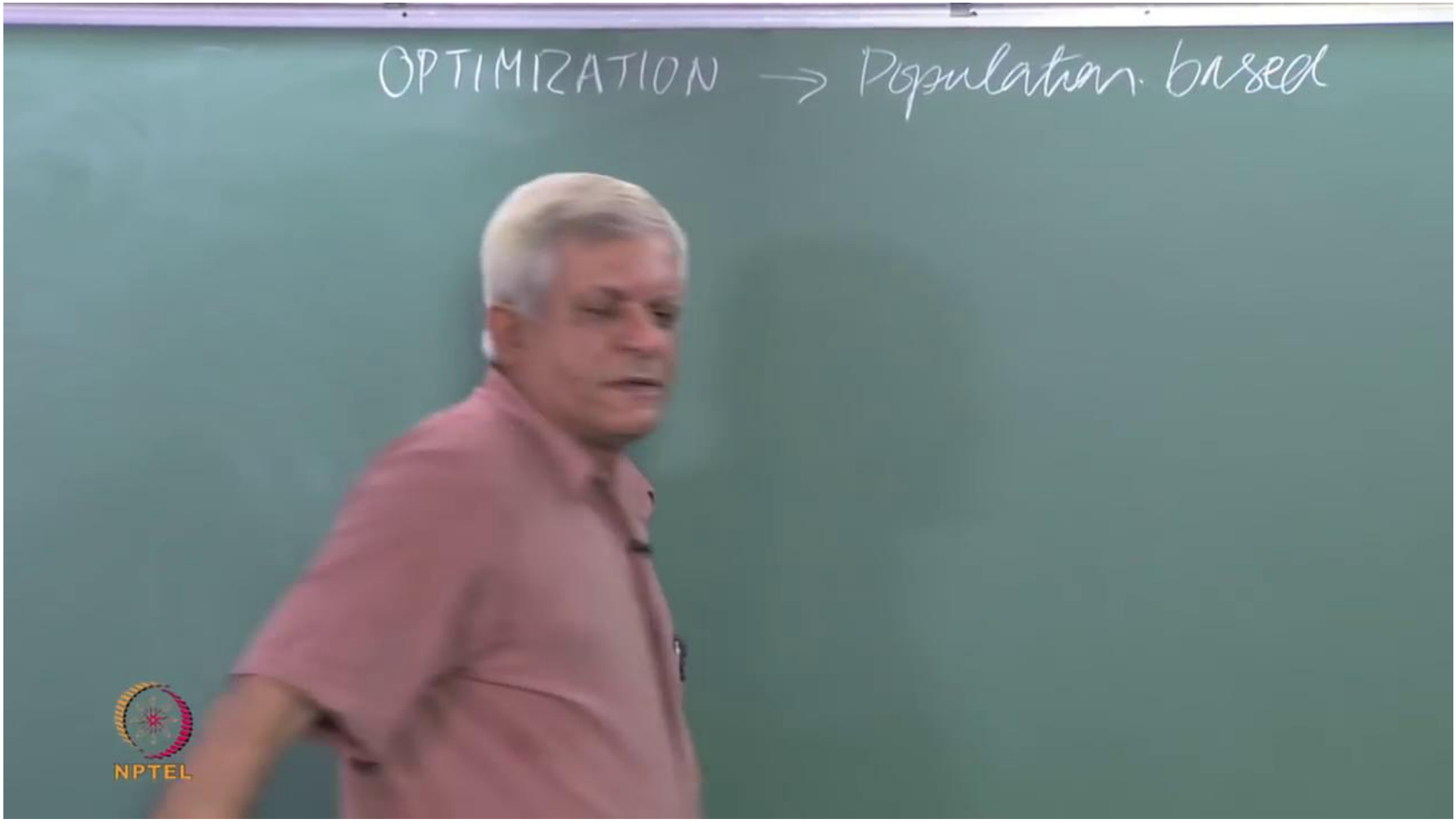
Video Lectures On

Artificial Intelligence

Lecture 16

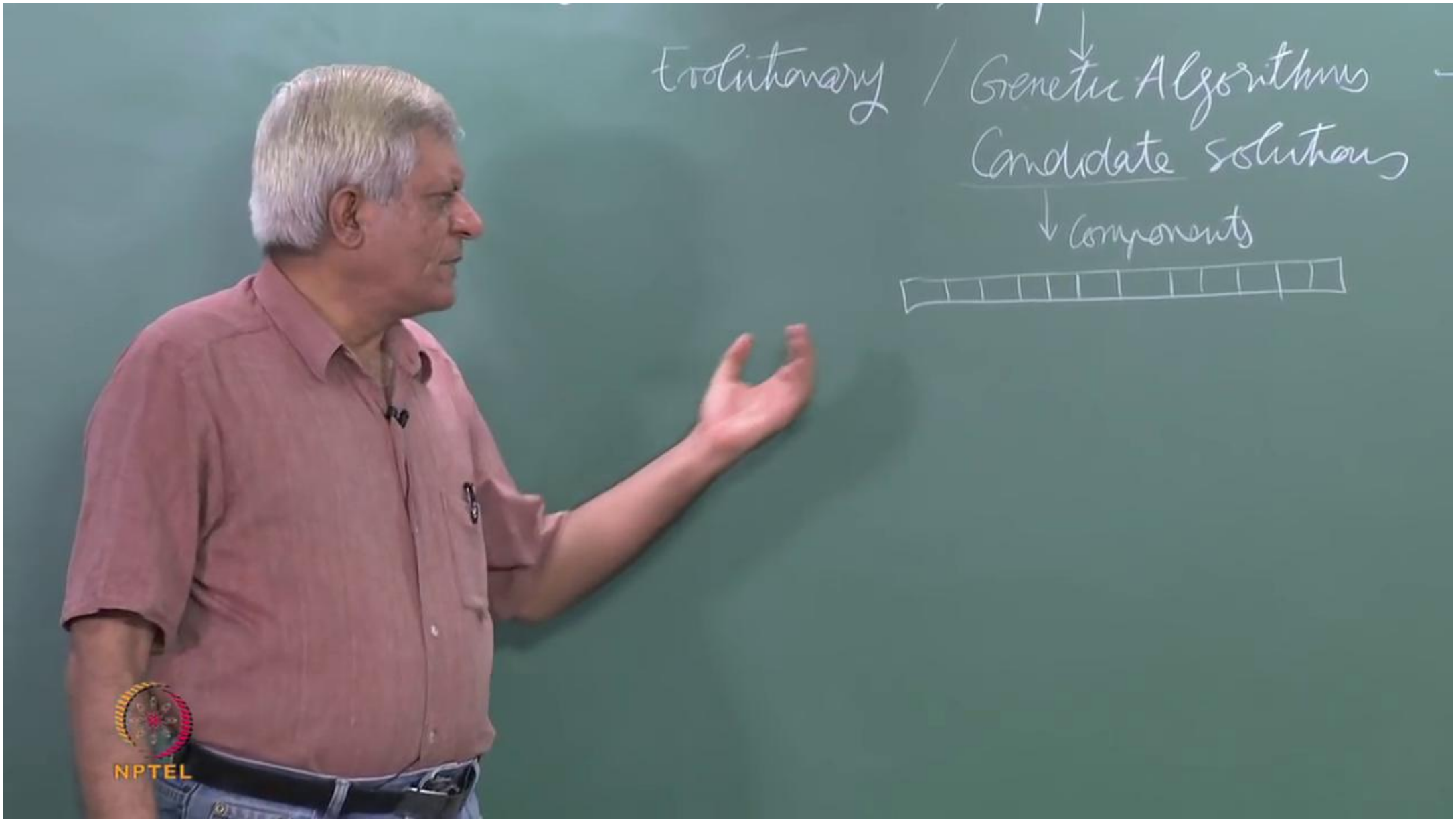
Population Based Methods for Optimization

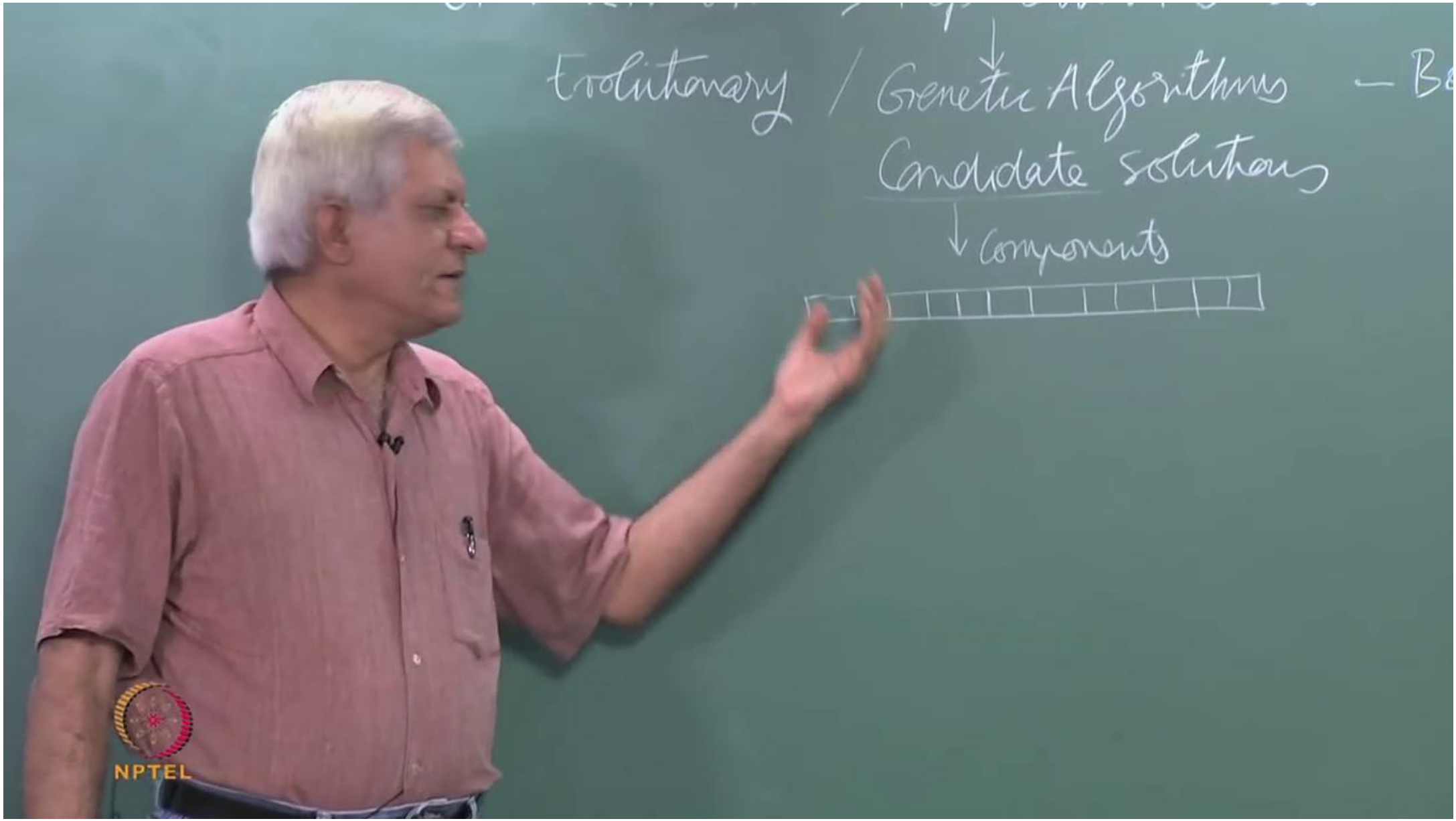
Prof. Deepak Khemani
Department of Computer Science
IIT Madras

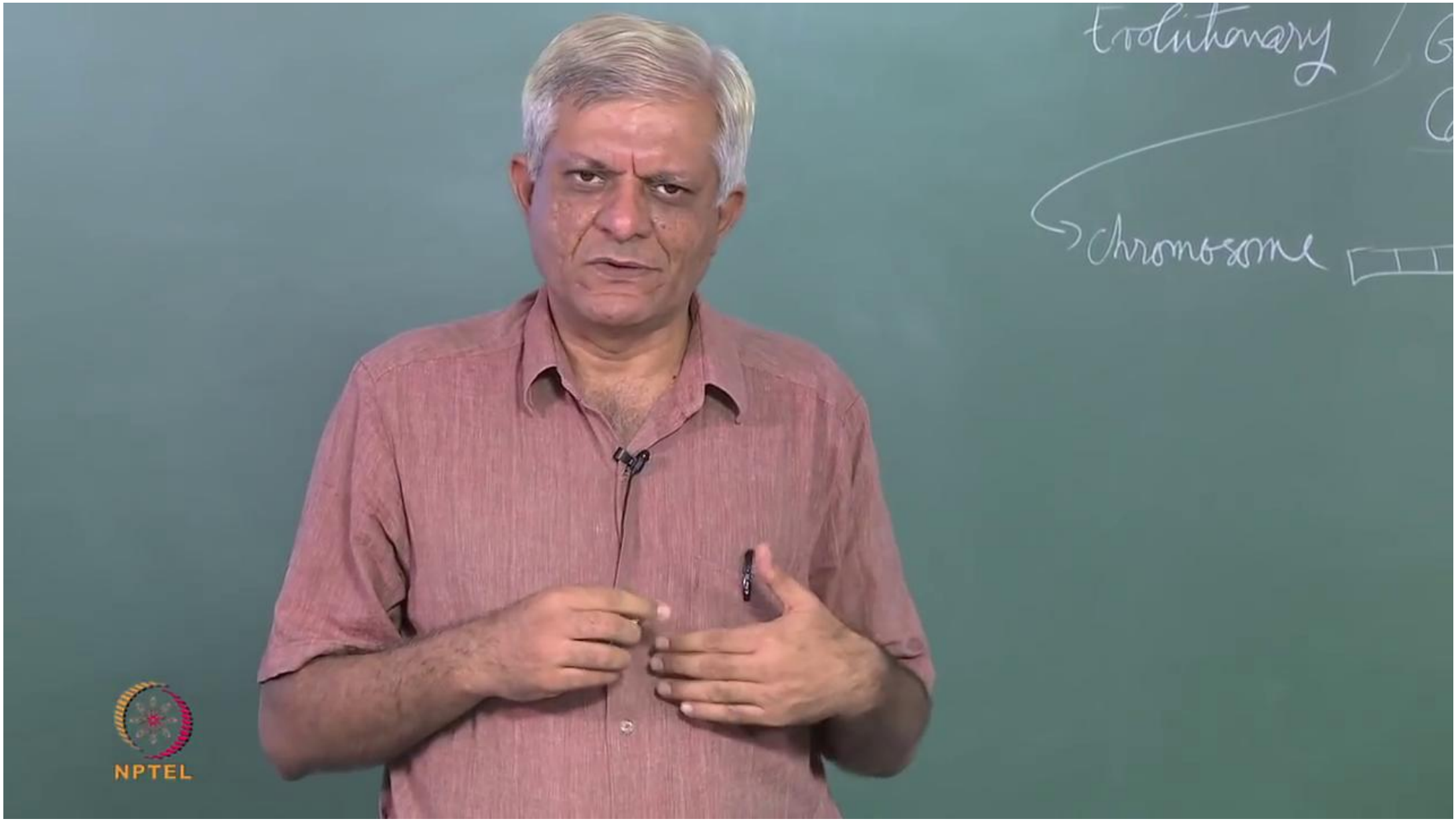

OPTIMIZATION → Population based
NPTEL

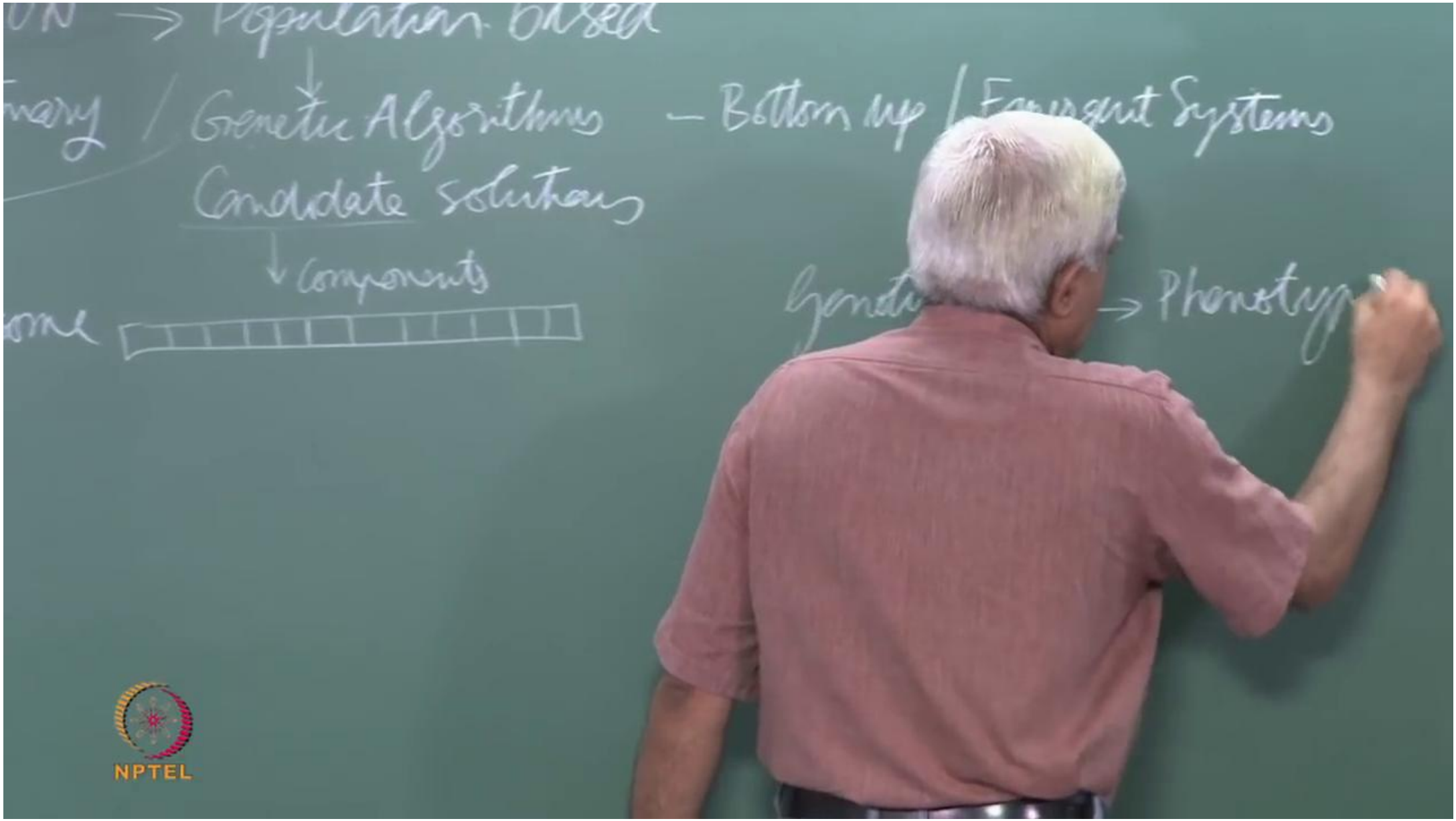
OPTIMIZATION → Population based

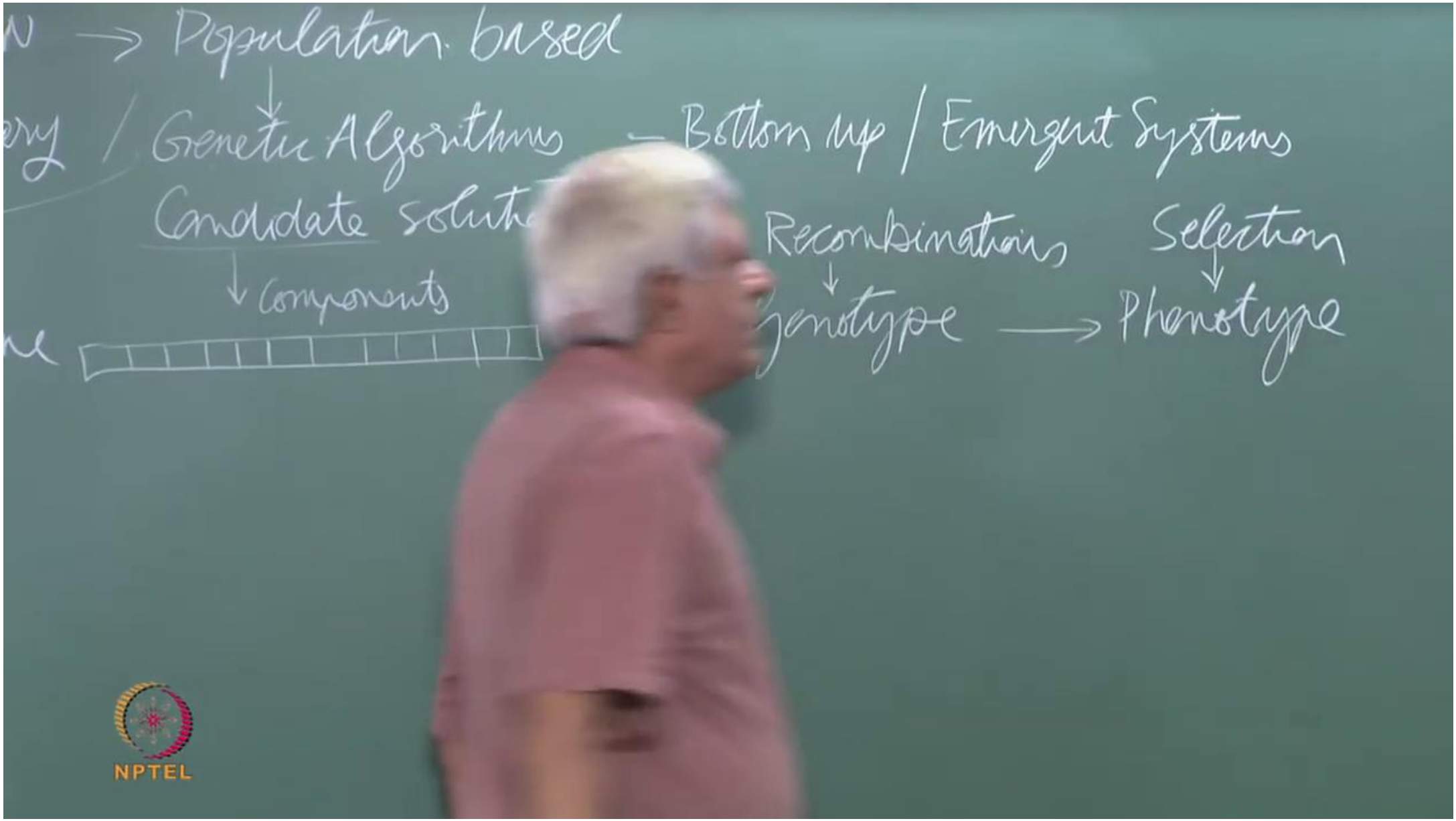
Evolutionary / Genetic Algorithms — Bottom up / Emergent Systems


Evolutionary / Genetic Algorithms
Candidate solutions
Components
NPTEL


Evolutionary / Genetic Algorithms
Candidate solutions
Components
NPTEL

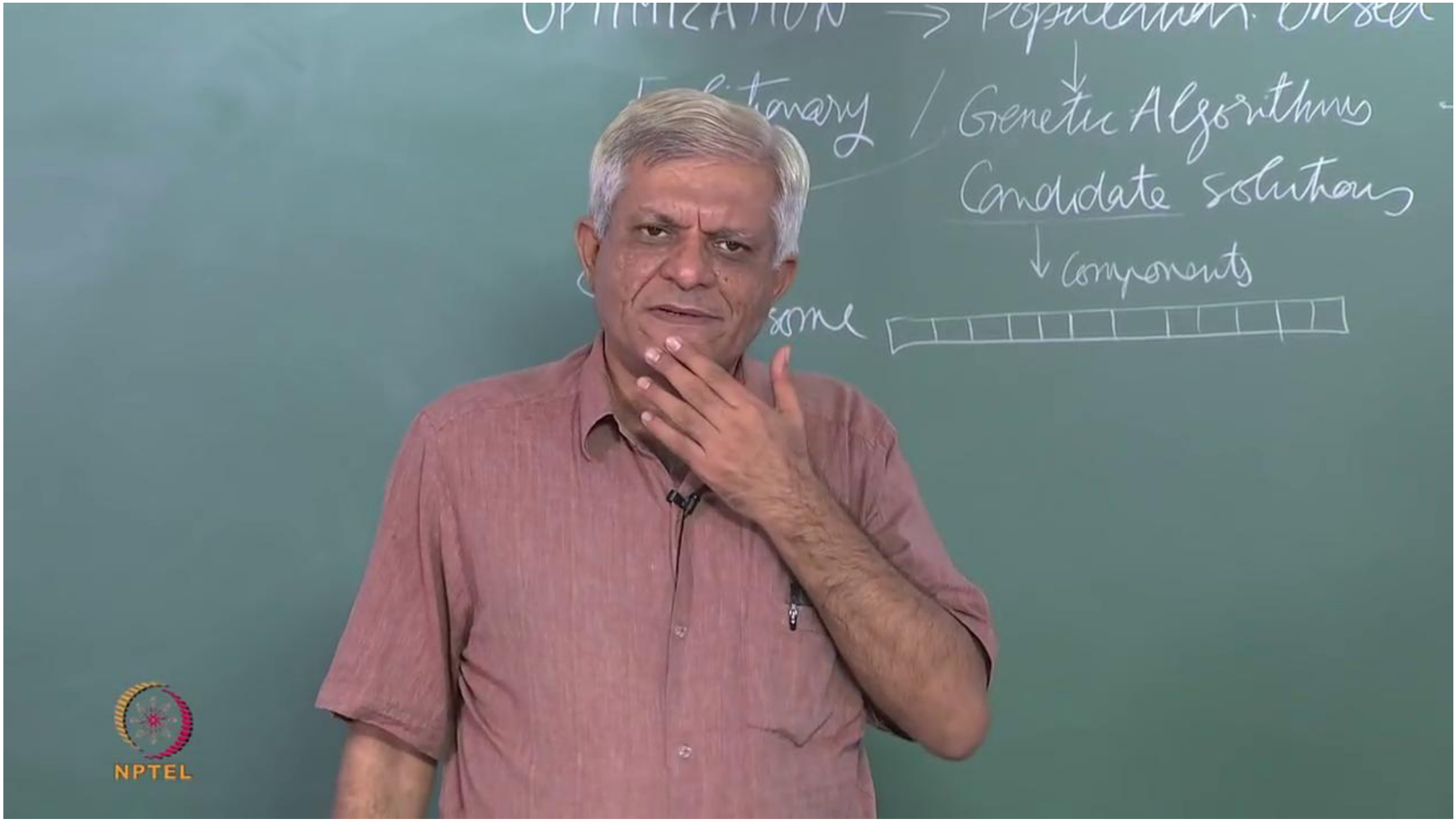

Evolutionary
Chromosome
NPTEL

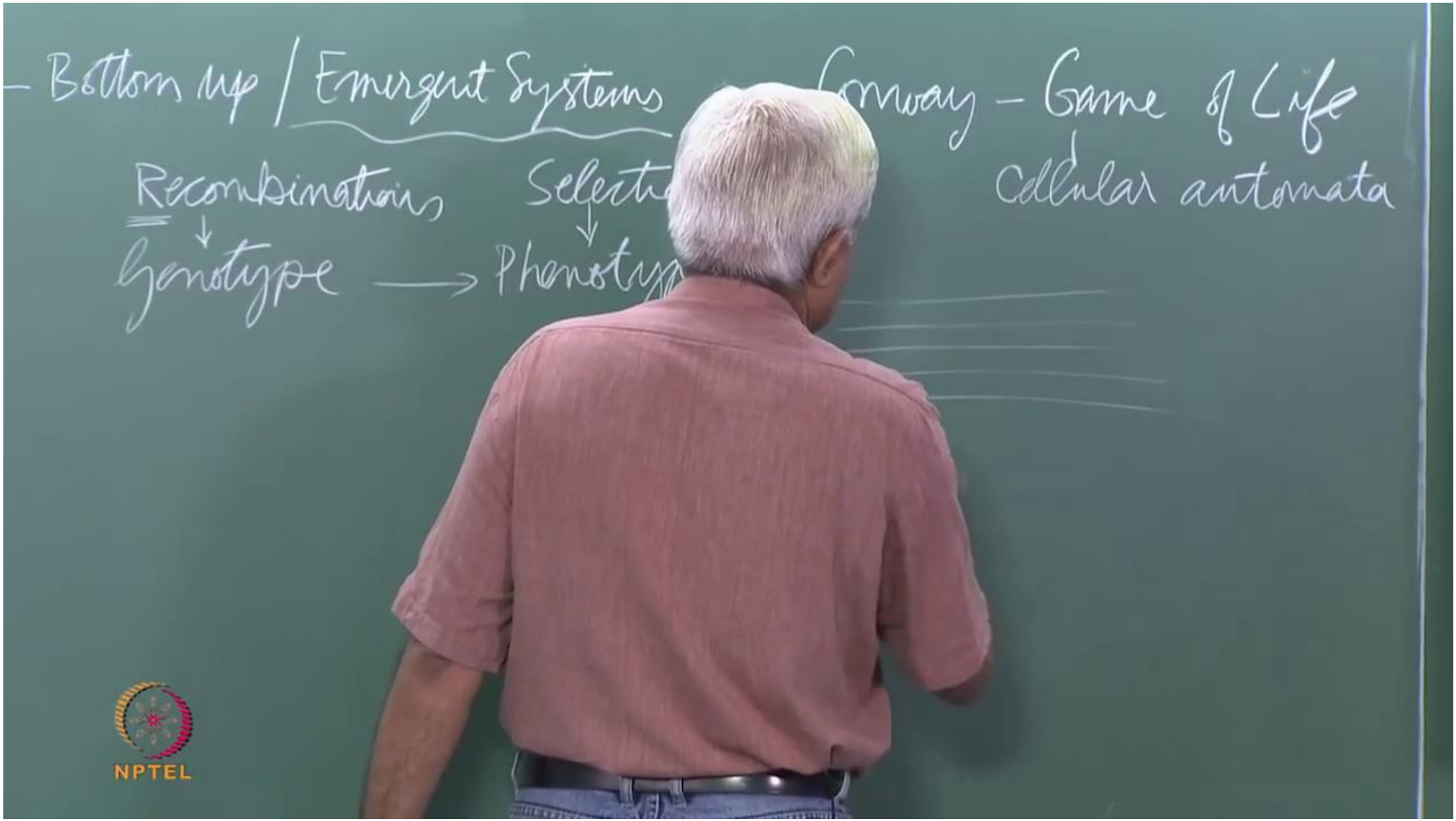

Genetic Algorithms
Candidate solutions
Components
Bottom up
NPTEL

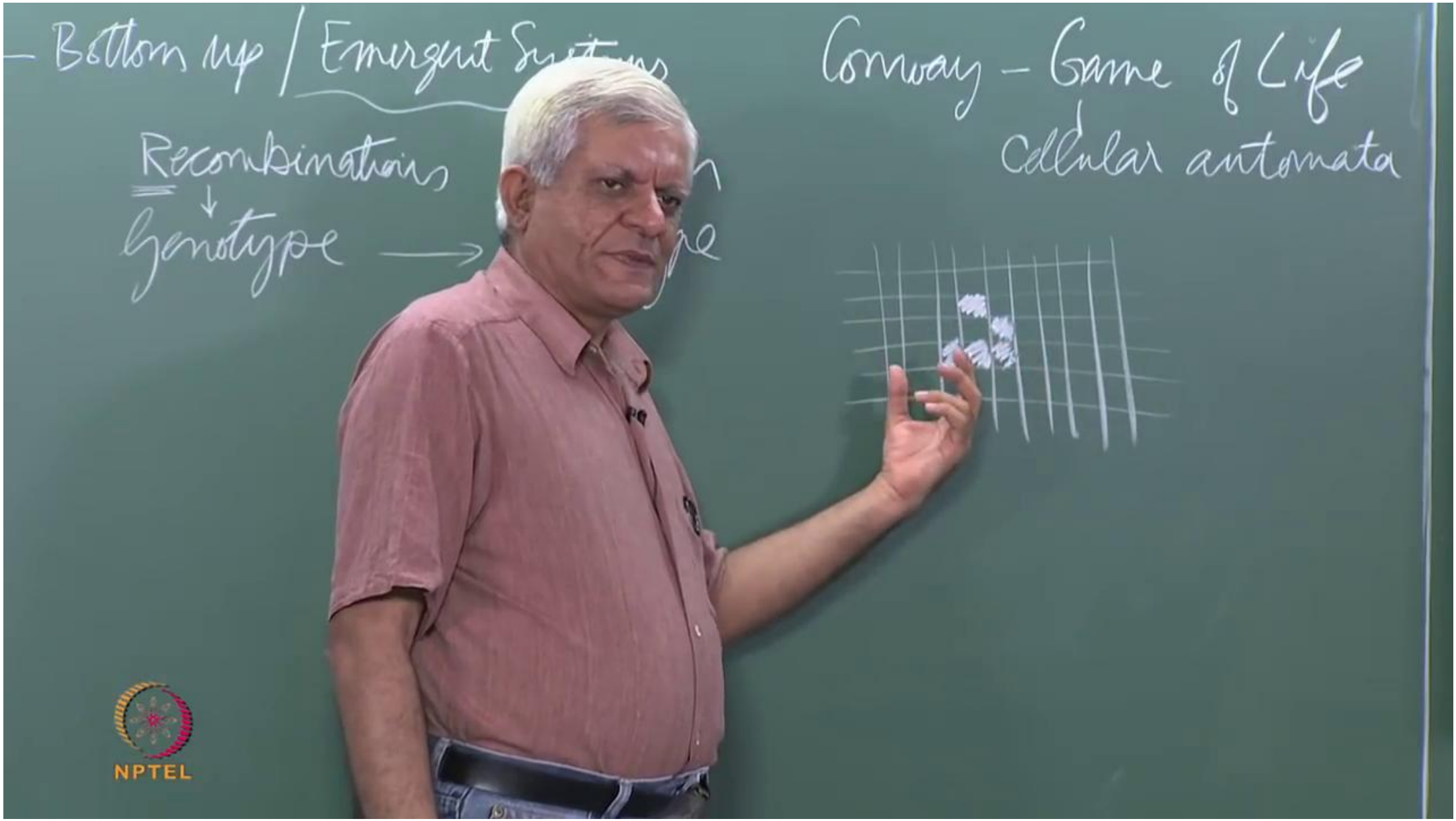

Population based
Genetic Algorithms
Bottom up / Emergent Systems
Candidate
Components
Recombination
Selection
Genotype
Phenotype
NPTEL

Recombinations → Genotype → Phenotype ← Selection

Recombinations → Genotype → Phenotype ← Selection

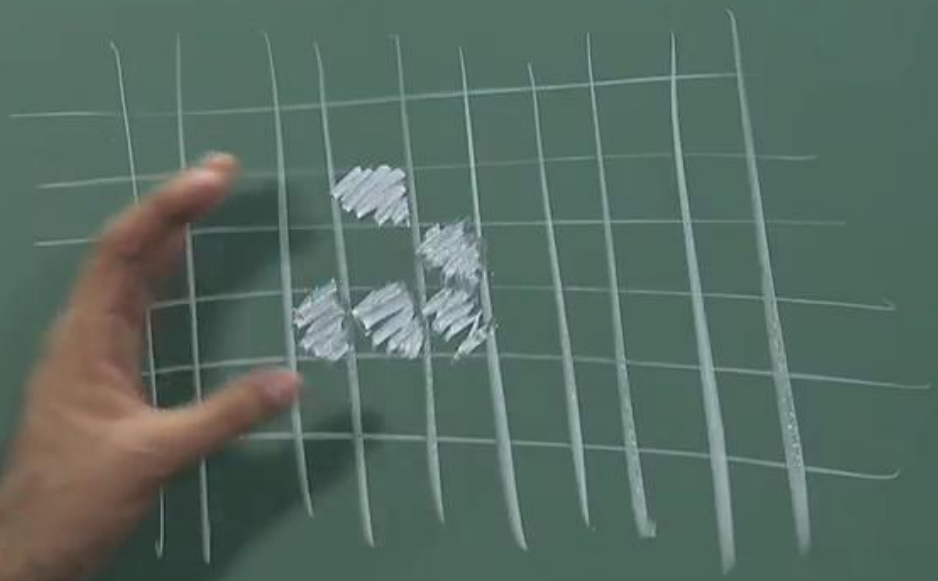

Genetic Algorithms
Candidate solutions
components


– Bottom up / Emergent Systems
Conway – Game of Life
Recombinations
Selecti
Genotype
Phenotyp
Cellular automata
NPTEL


– Bottom up / Emergent
Recombinations
Genotype
Conway – Game of Life
cellular automata
NPTEL

Conway – Game of Life

cellular automata



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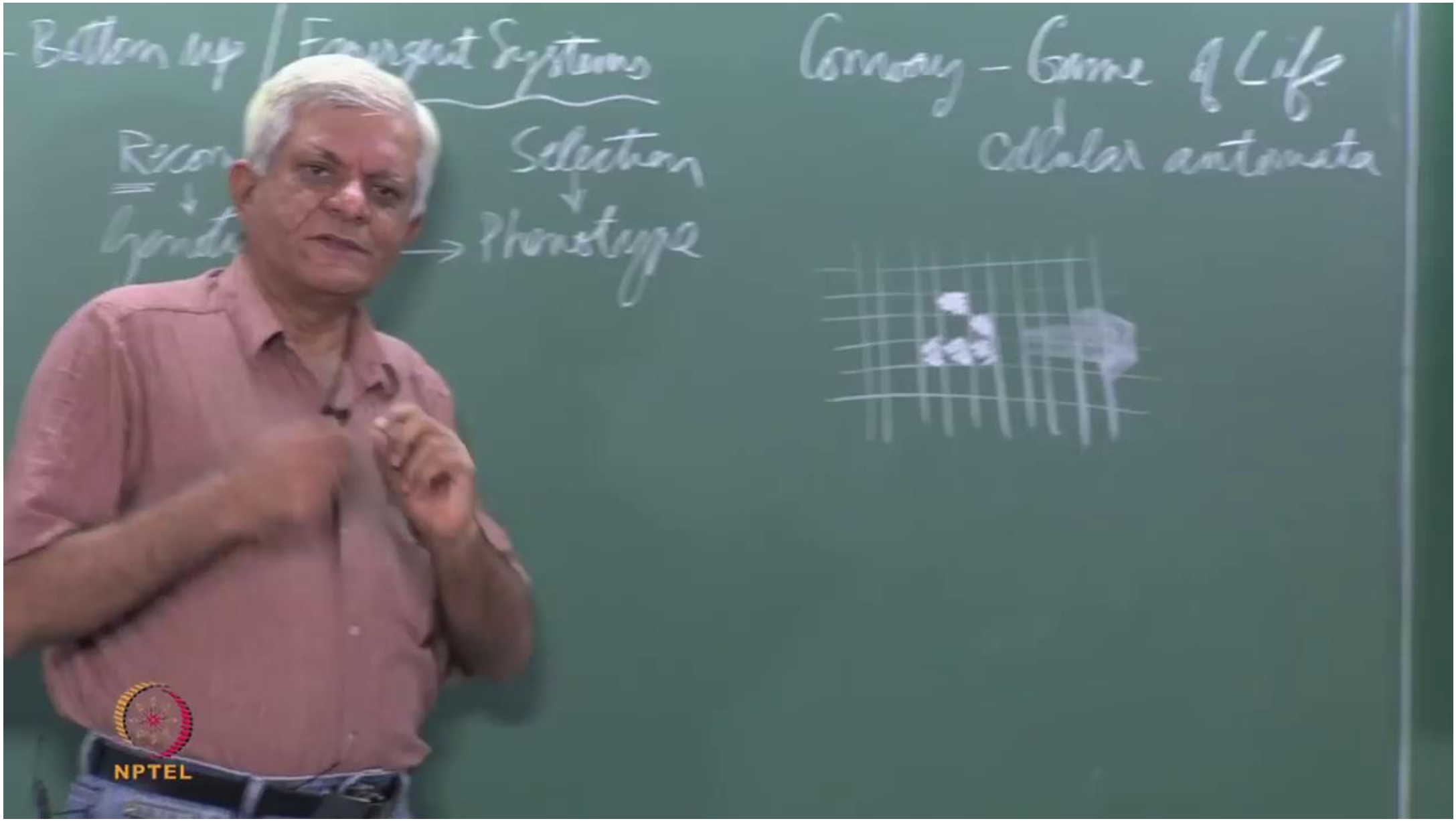
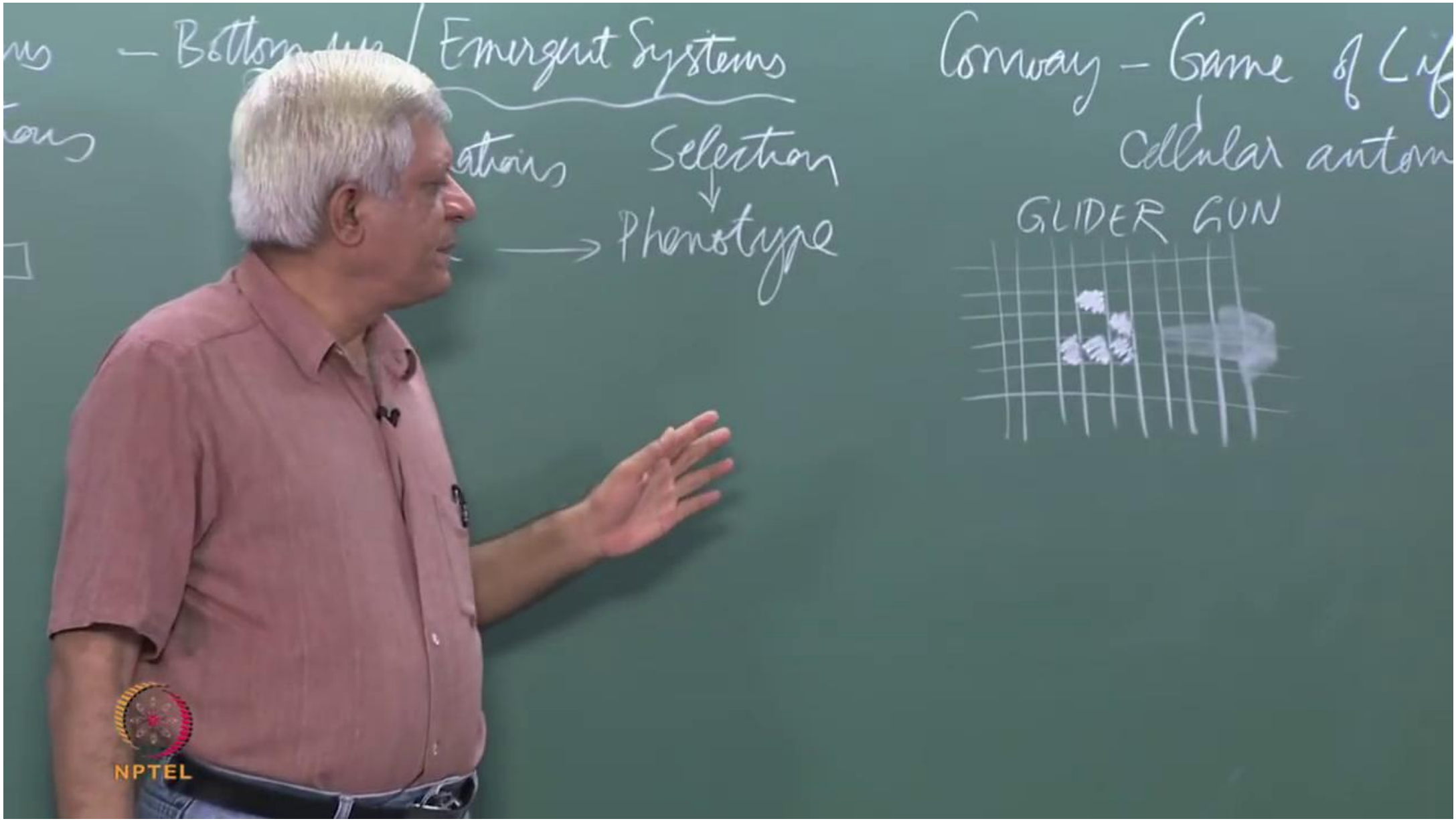
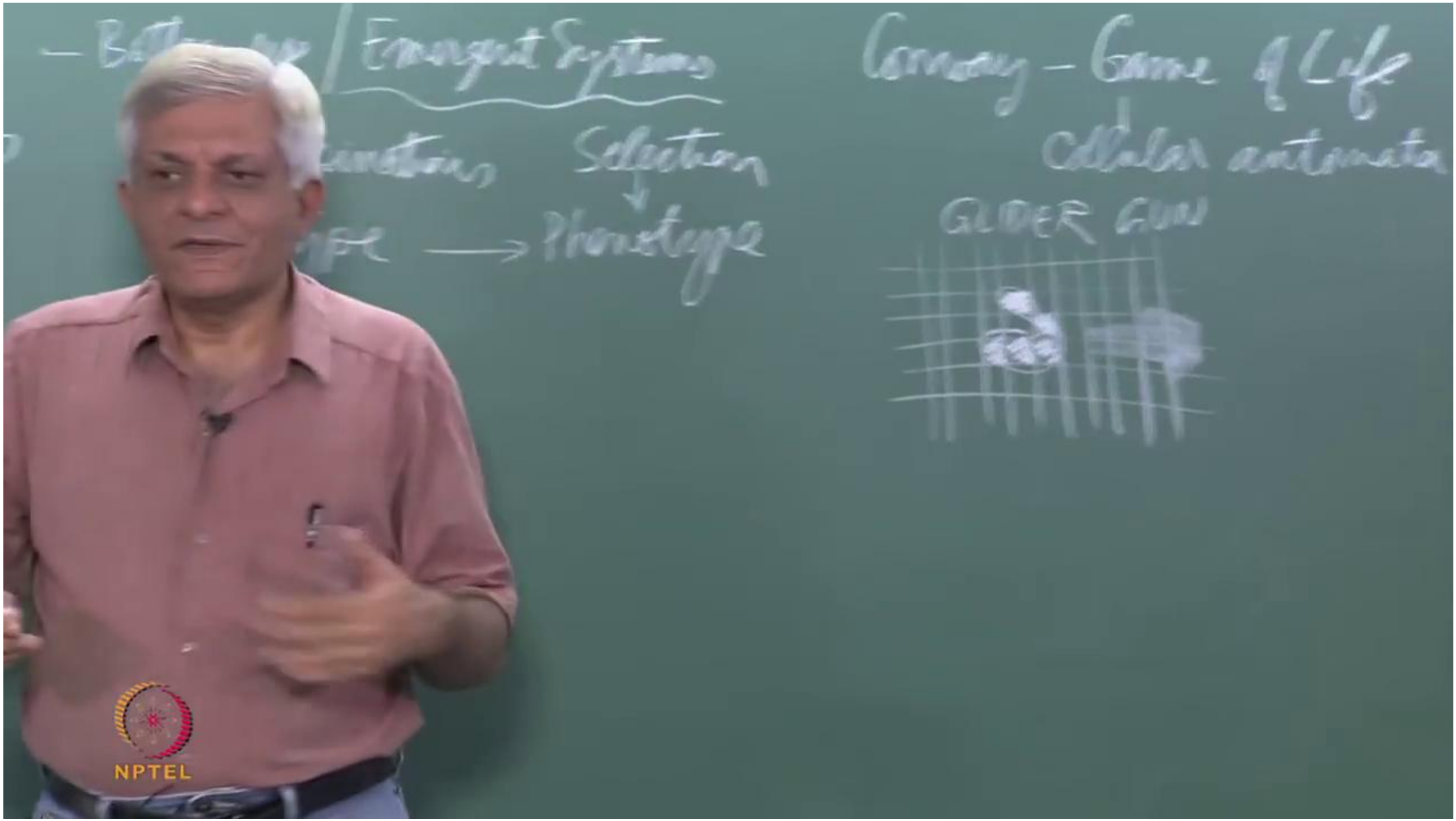
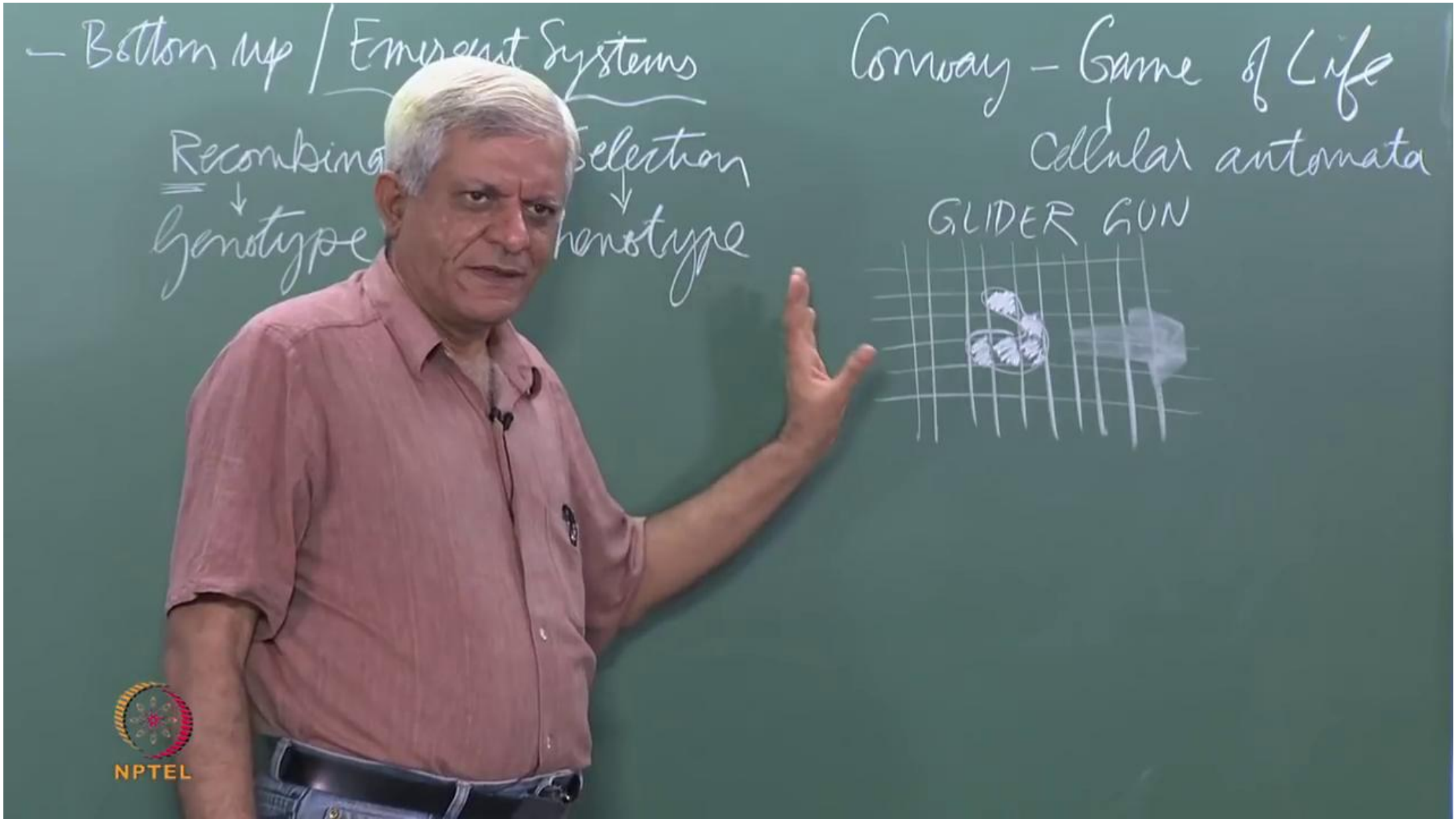
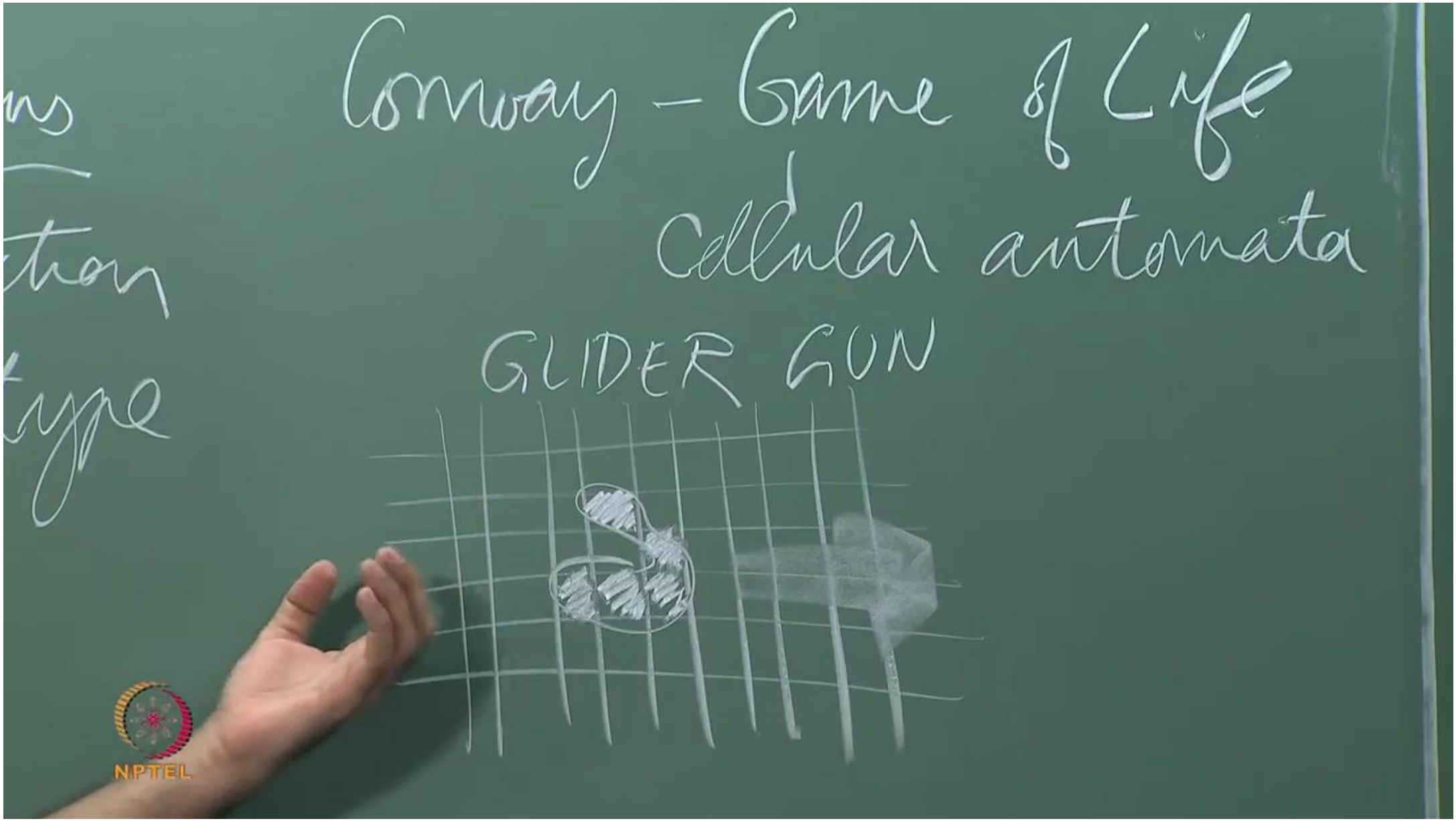
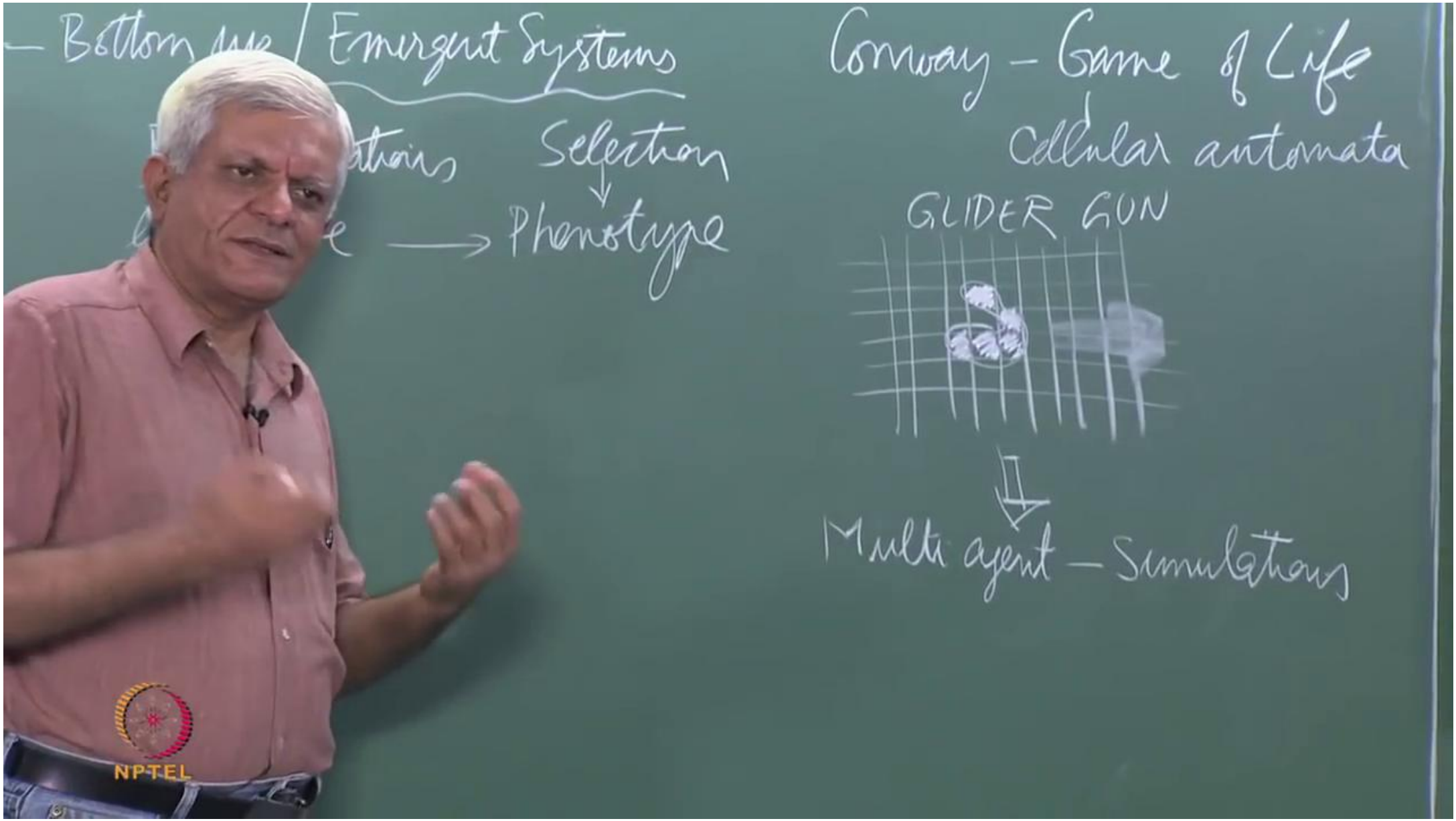

Bottom up
Emergent Systems
Recon
Selection
Phenotype
Conway – Game of Life
Cellular automata
NPTEL

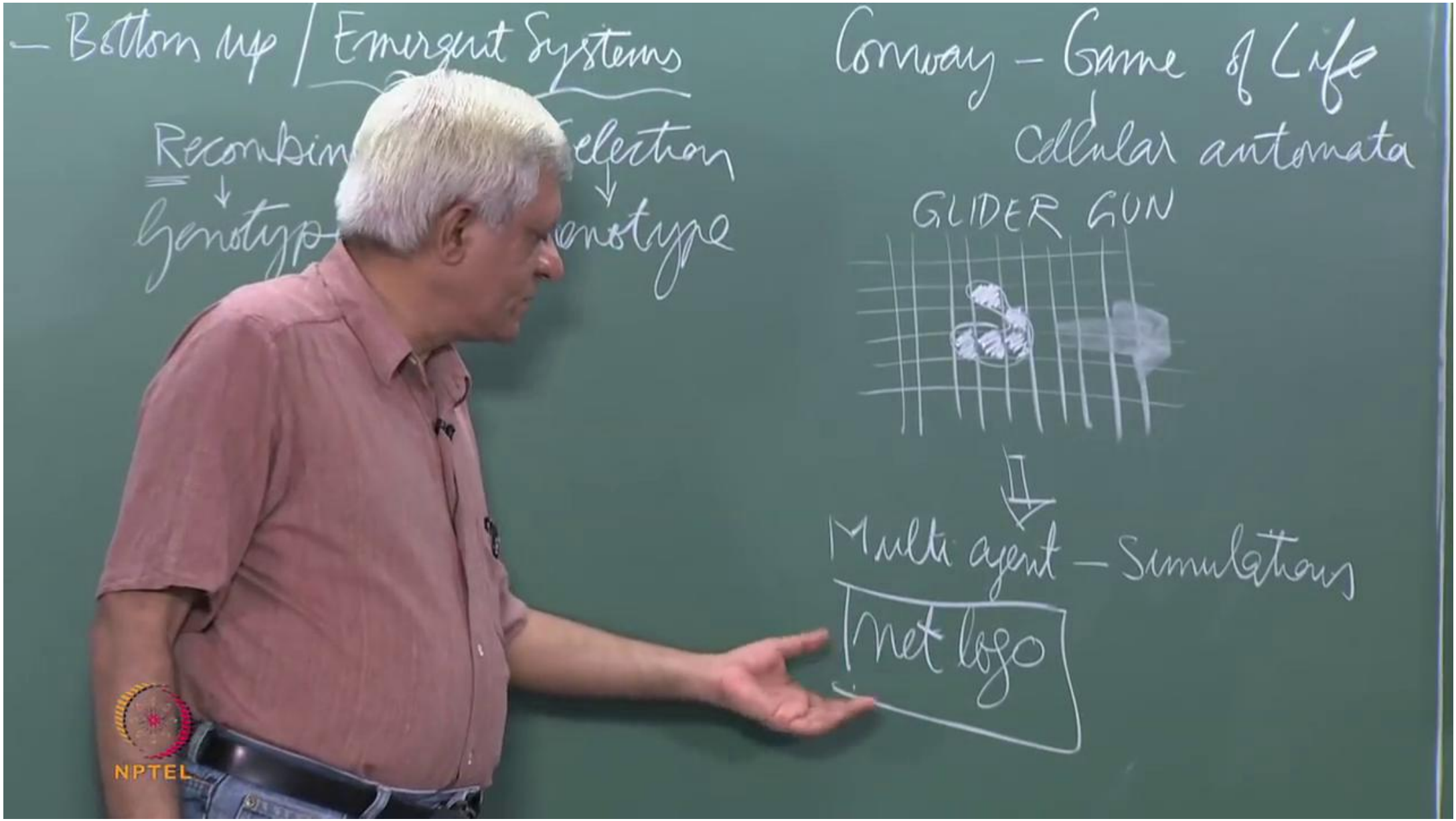

Emergent Systems
Selection
Phenotype
Conway – Game of Life
Cellular autom
GLIDER GUN
NPTEL

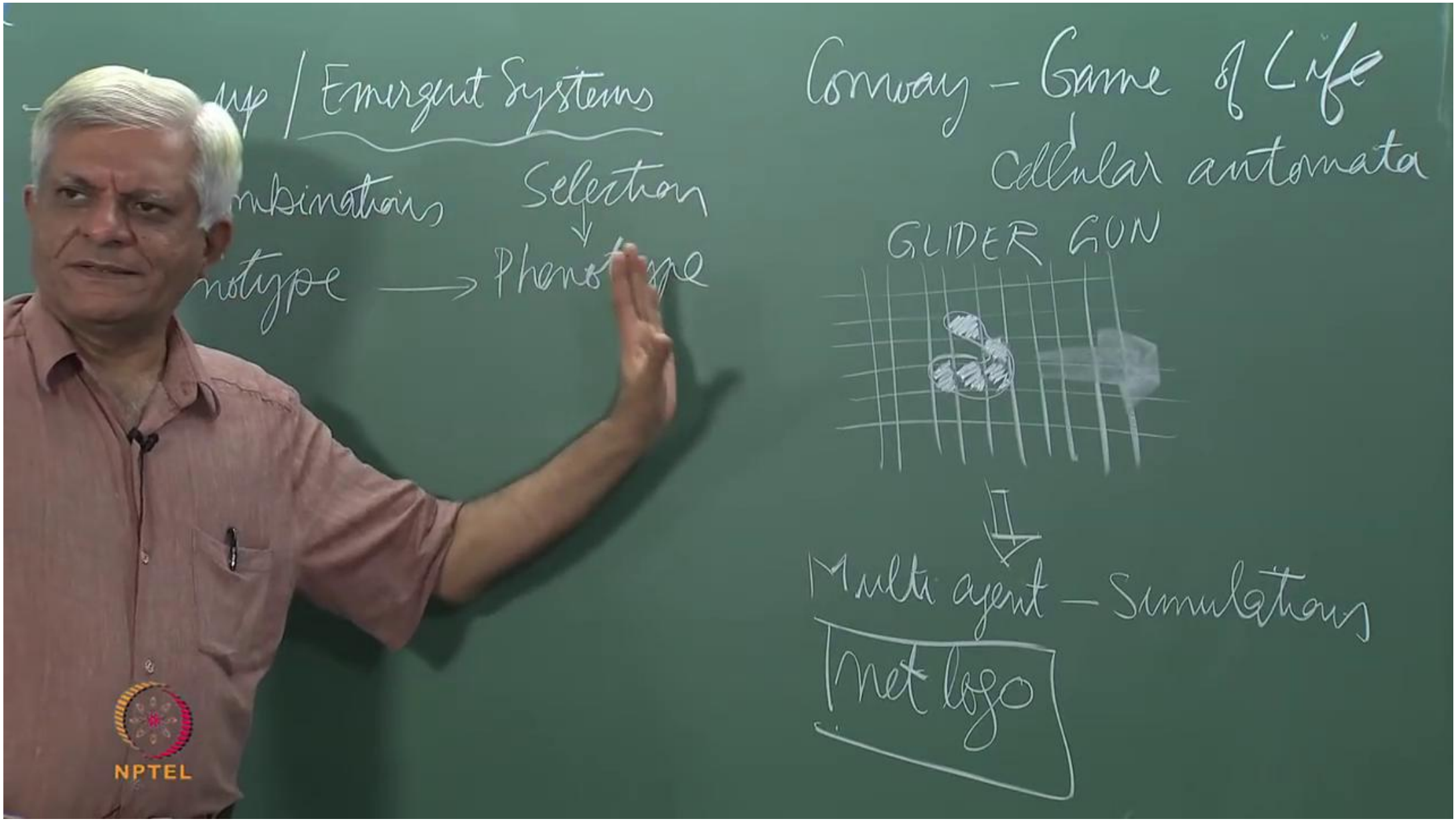

Emergent Systems
Selection
Prototype
Game of Life
Cellular automata
GLIDER GUN

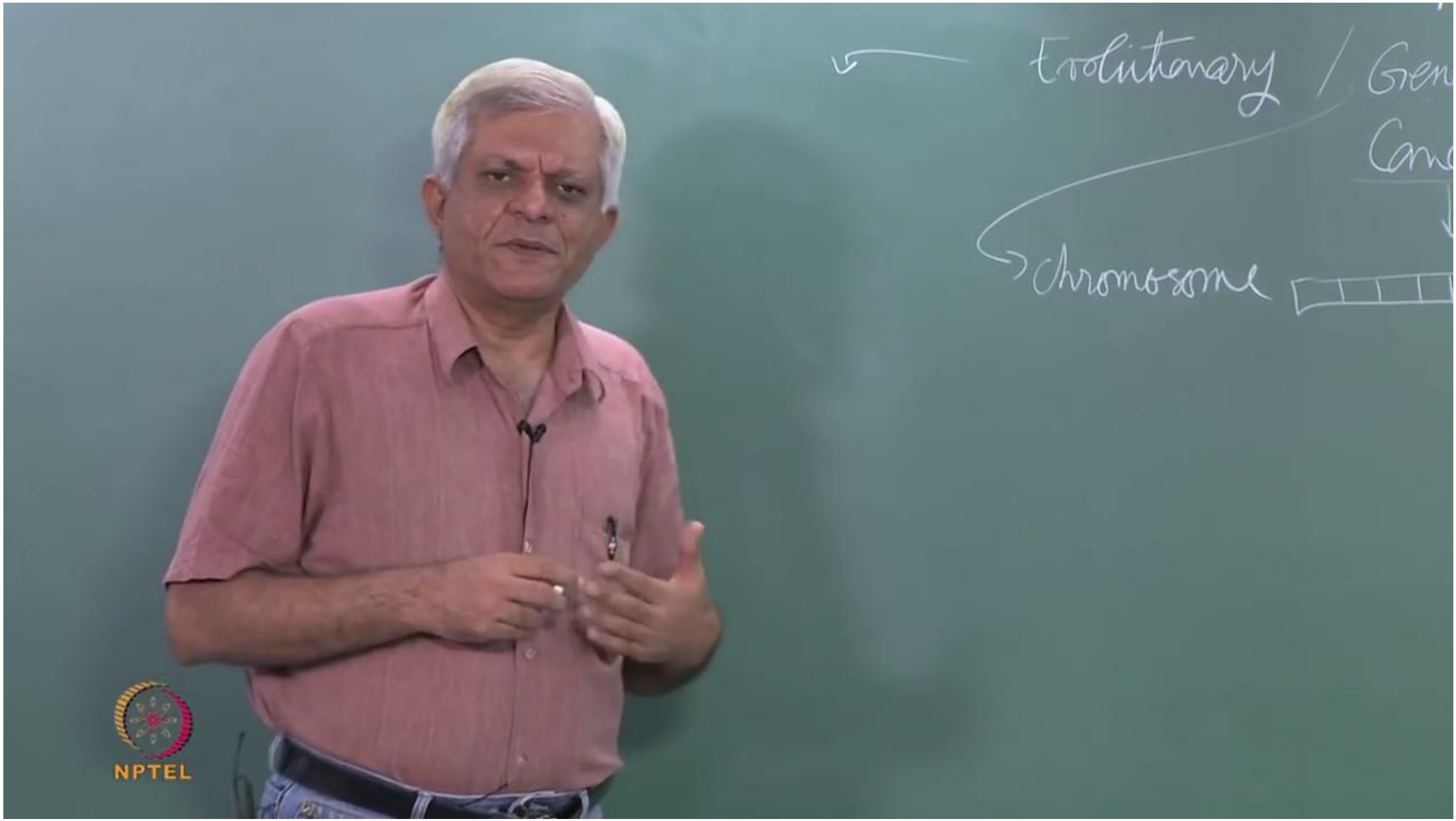

– Bottom up / Emergent Systems
Recombina
Genotype
Selection
Phenotype
Conway – Game of Life
Cellular automata
GLIDER GUN
NPTEL

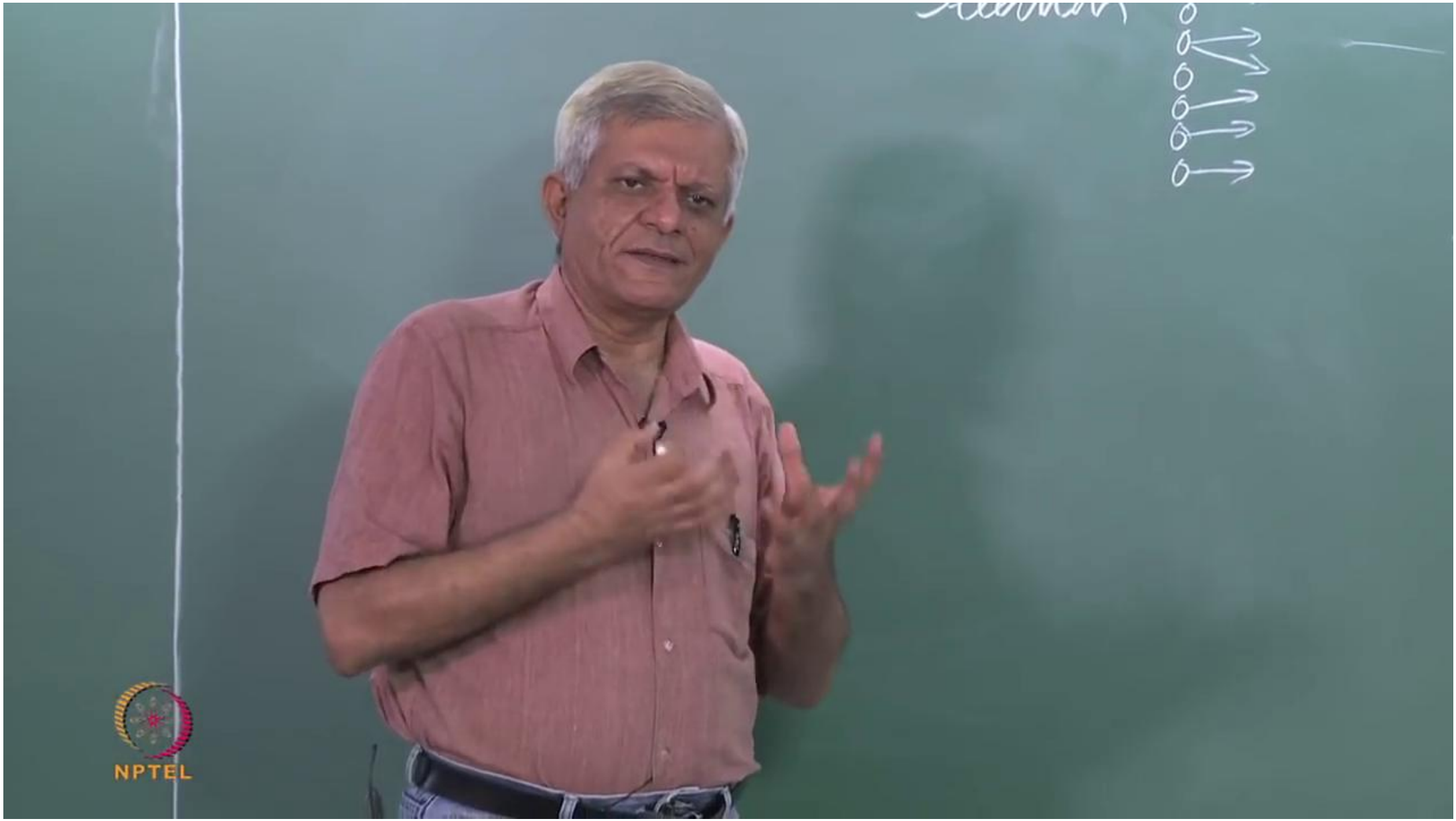

Conway – Game of Life
cellular automata
GLIDER GUN
NPTEL


Emergent Systems
Selection
Phenotype
Conway – Game of Life
Cellular automata
GLIDER GUN
Multi agent – Simulations
NPTEL


– Bottom up / Emergent Systems
Conway – Game of Life
Cellular automata
GLIDER GUN
Multi agent – Simulations
net logo
NPTEL


Emergent Systems
Selection
Phenotype
Conway – Game of Life
Cellular automata
GLIDER GUN
Multi agent – Simulations
Net logo
NPTEL


Evolutionary
Chromosome
NPTEL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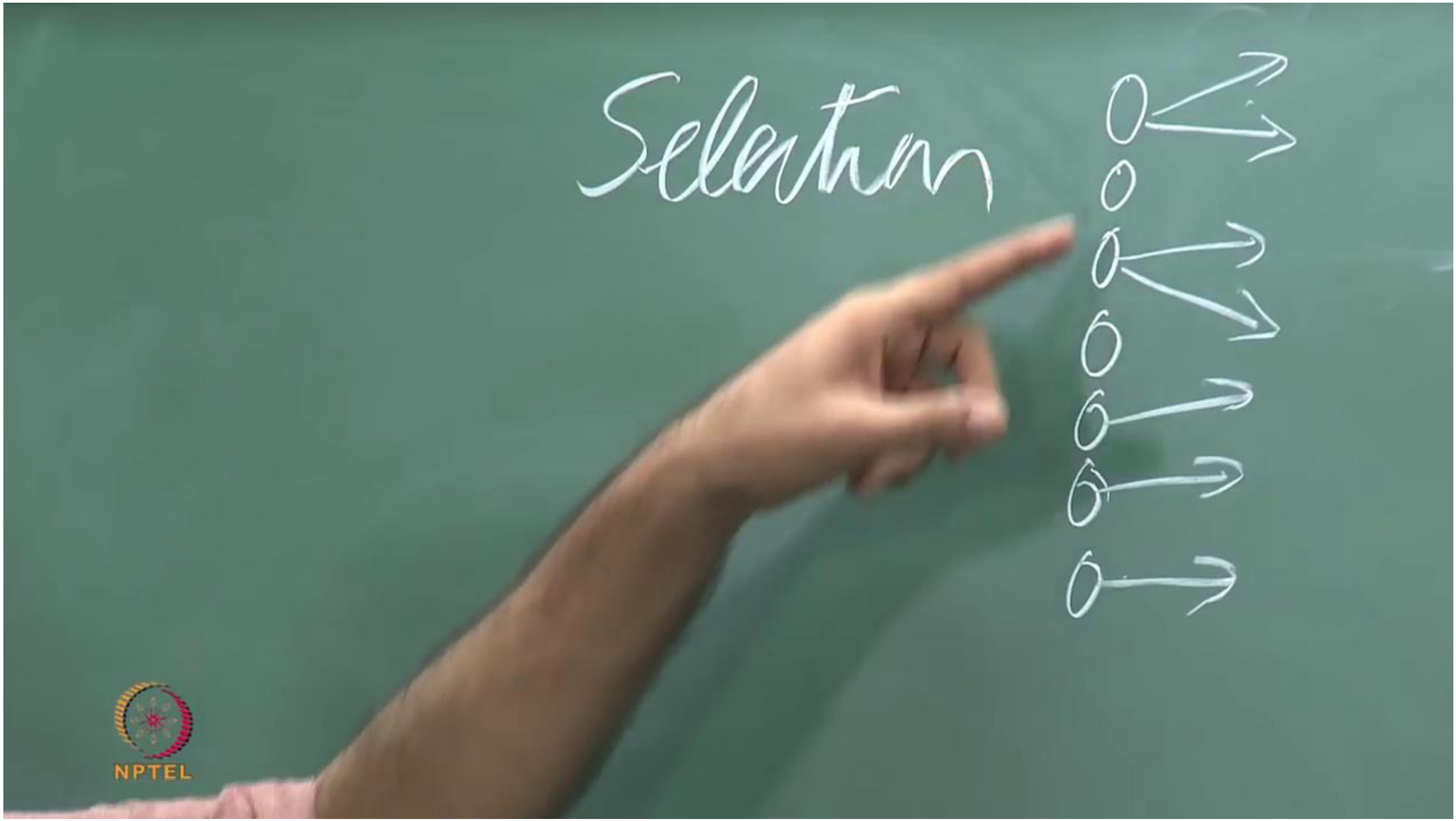
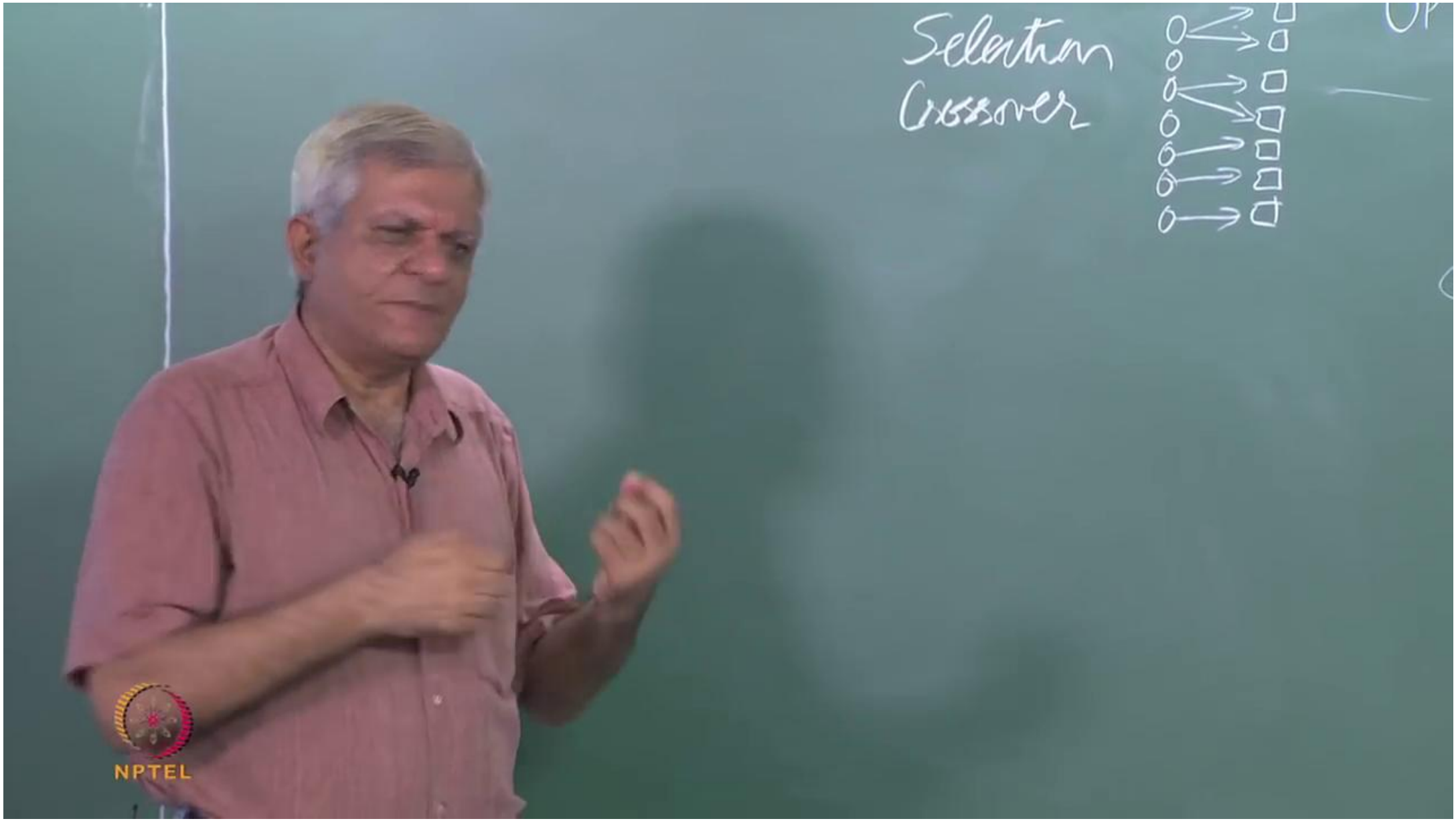
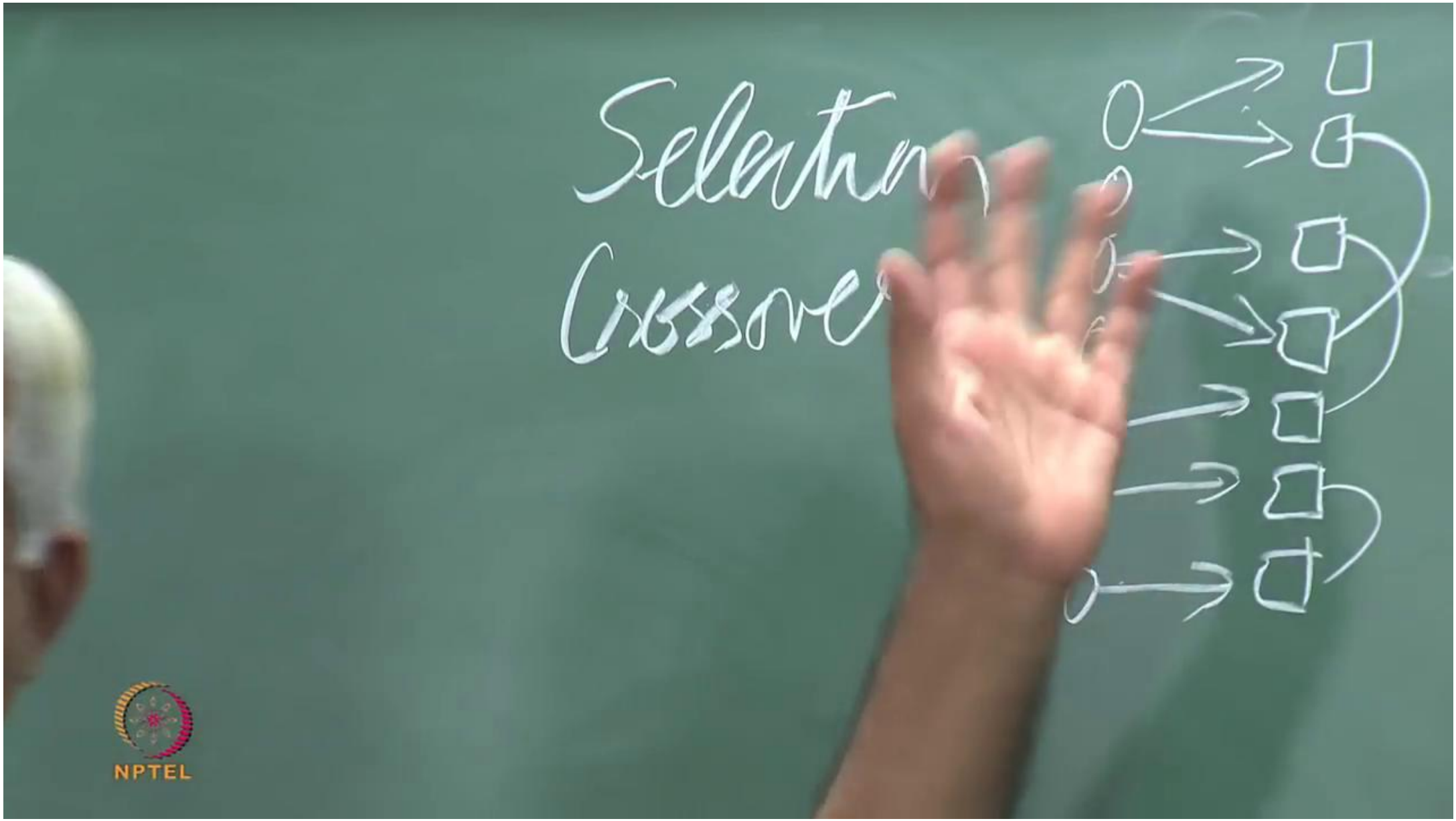
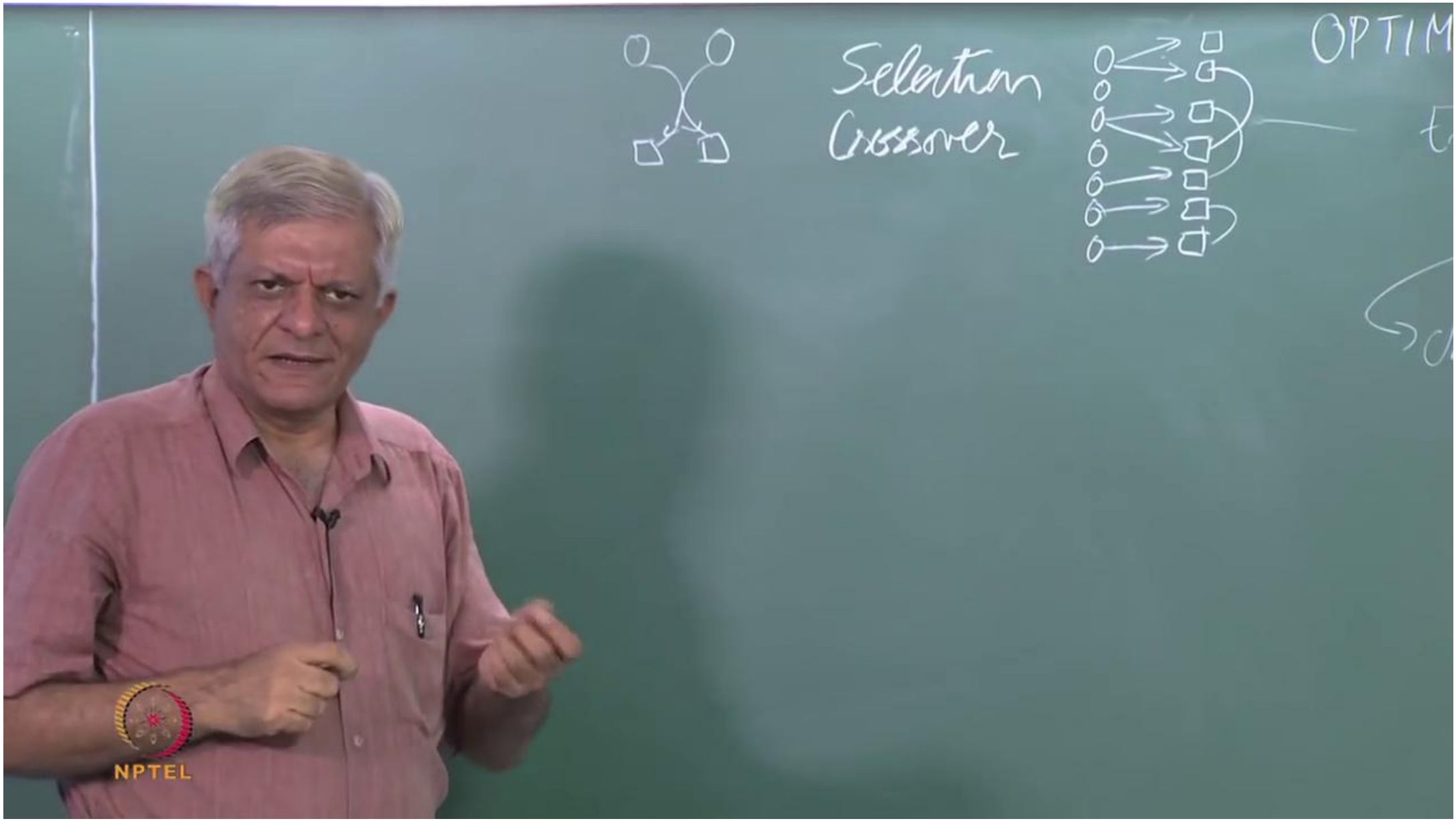
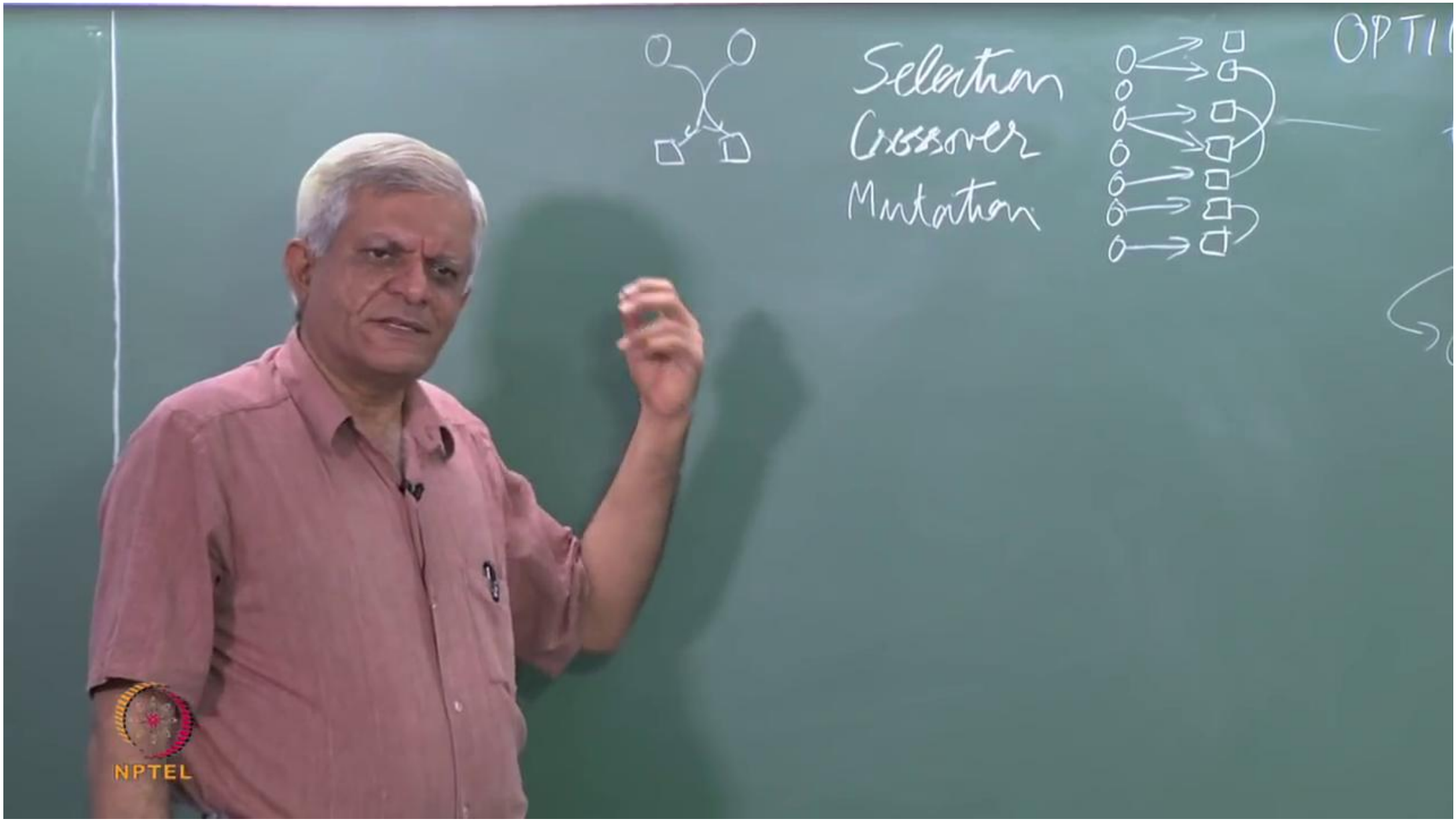
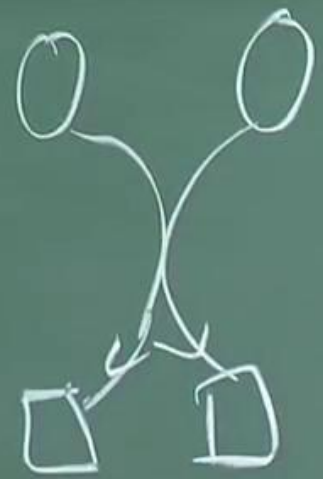

Selection


Selection
Crossover
NPTEL


Selection
Crossover
NPTEL


Selection
Crossover
OPTIM
NPTEL


Selection
Crossover
Mutation
NPTEL

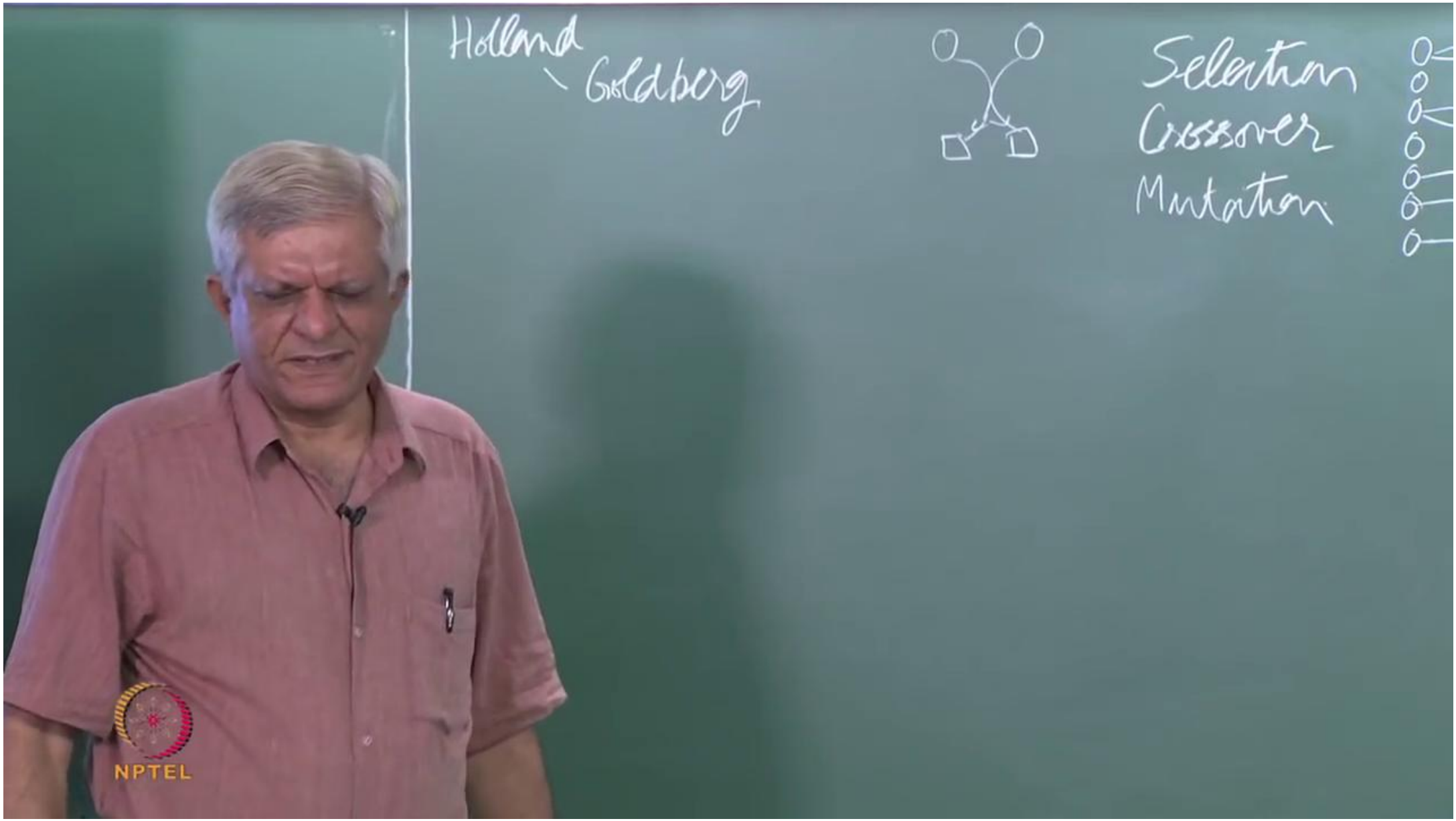


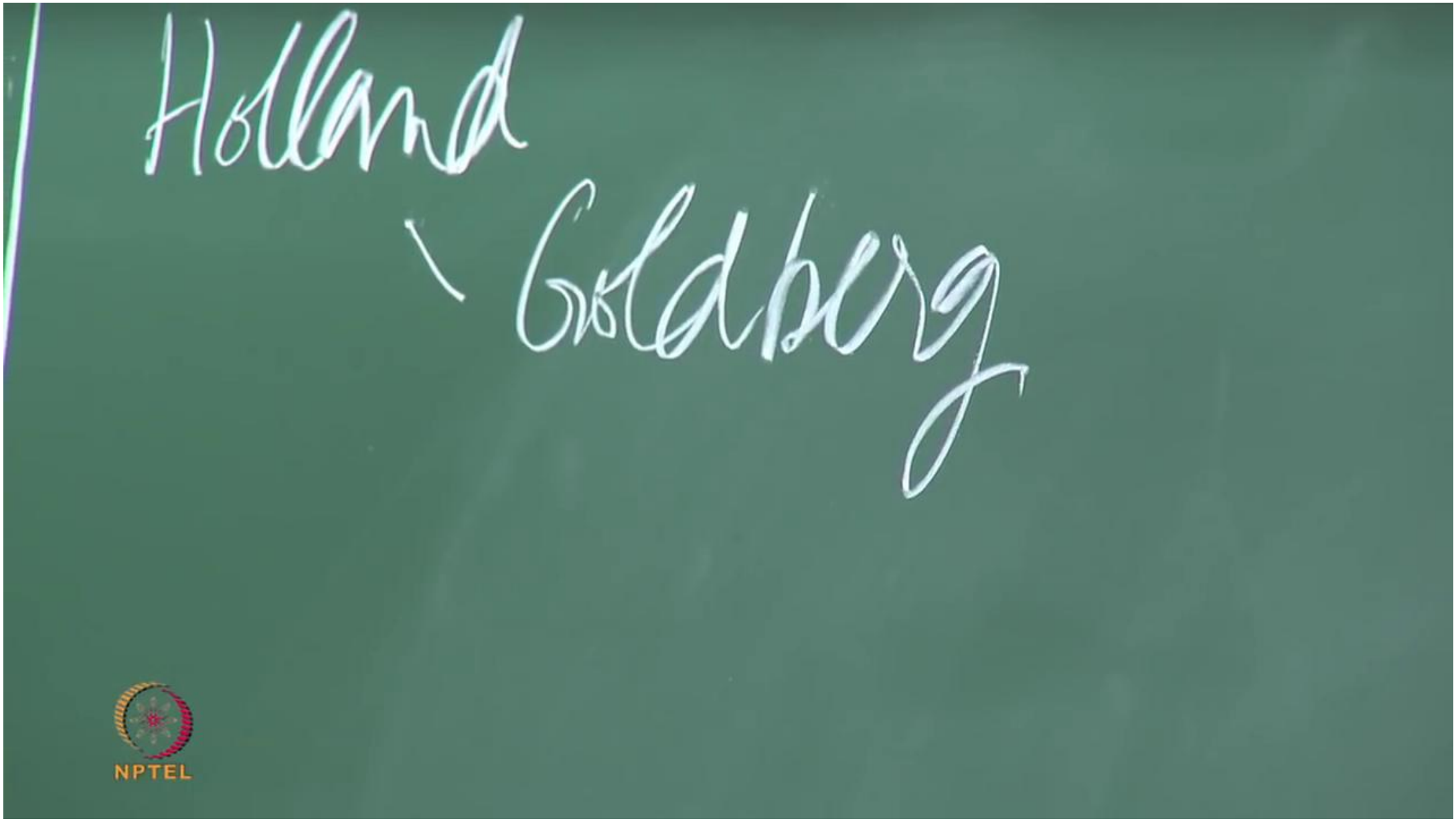
Selection

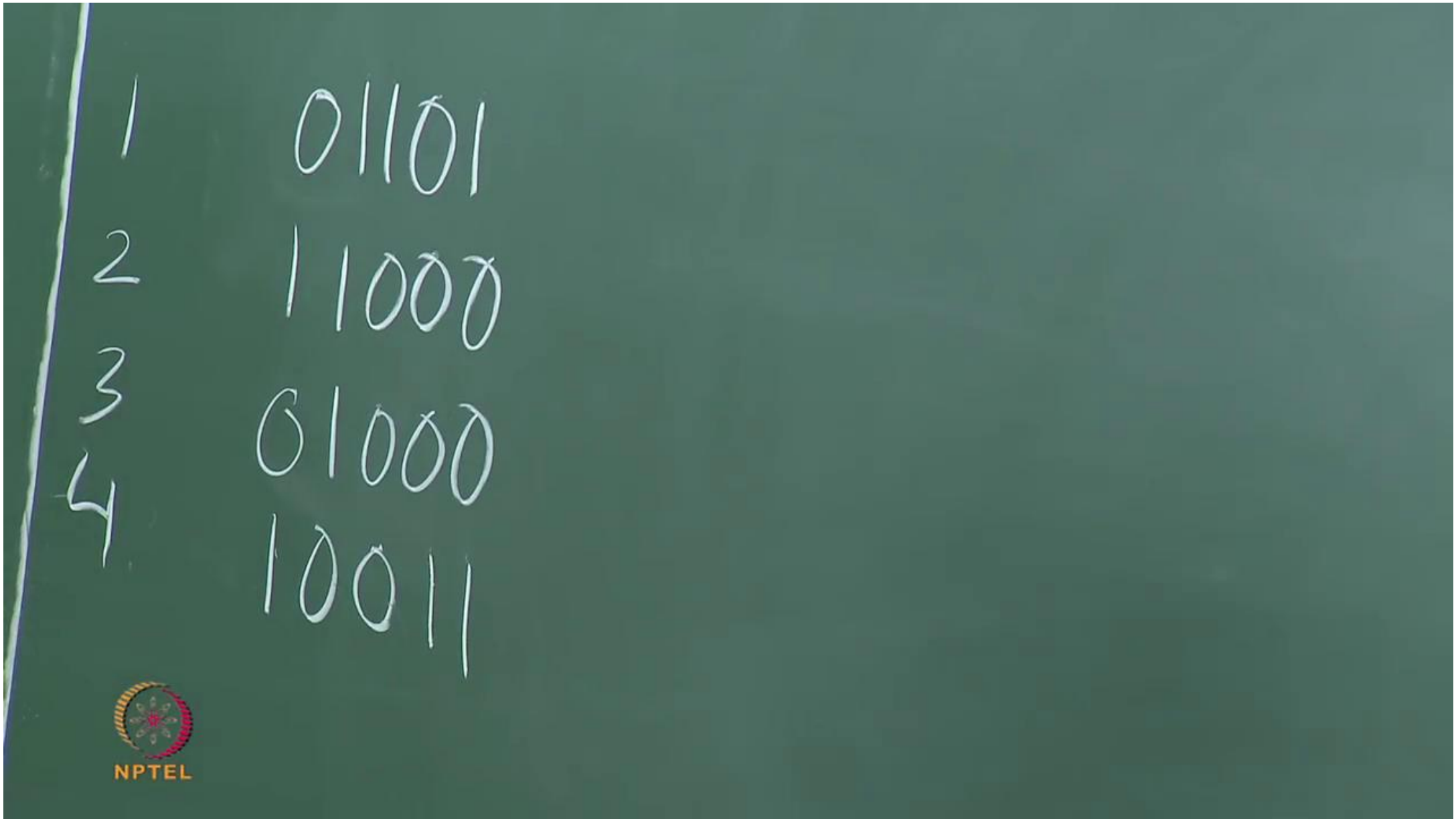
Crossover

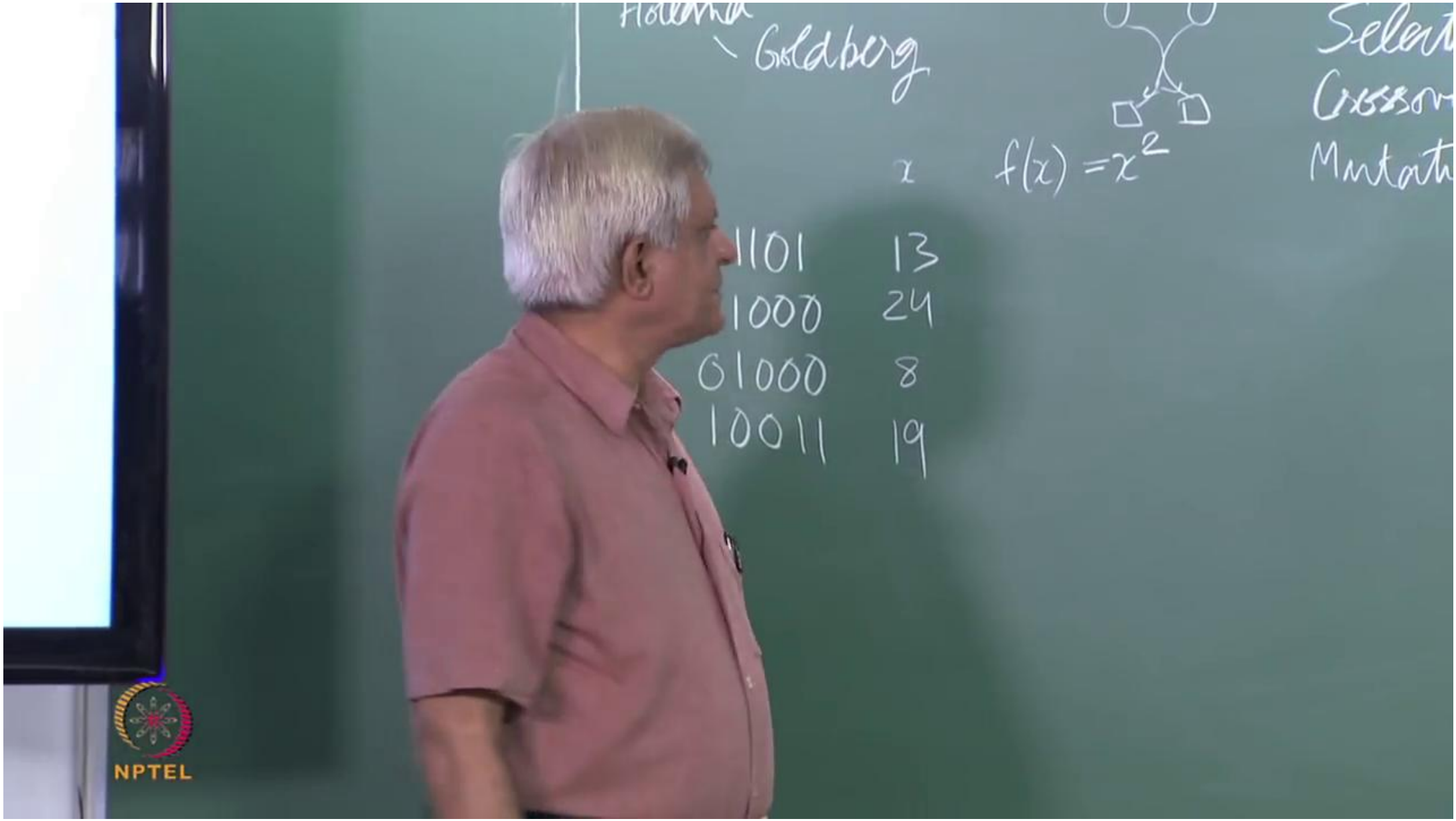
Mutation

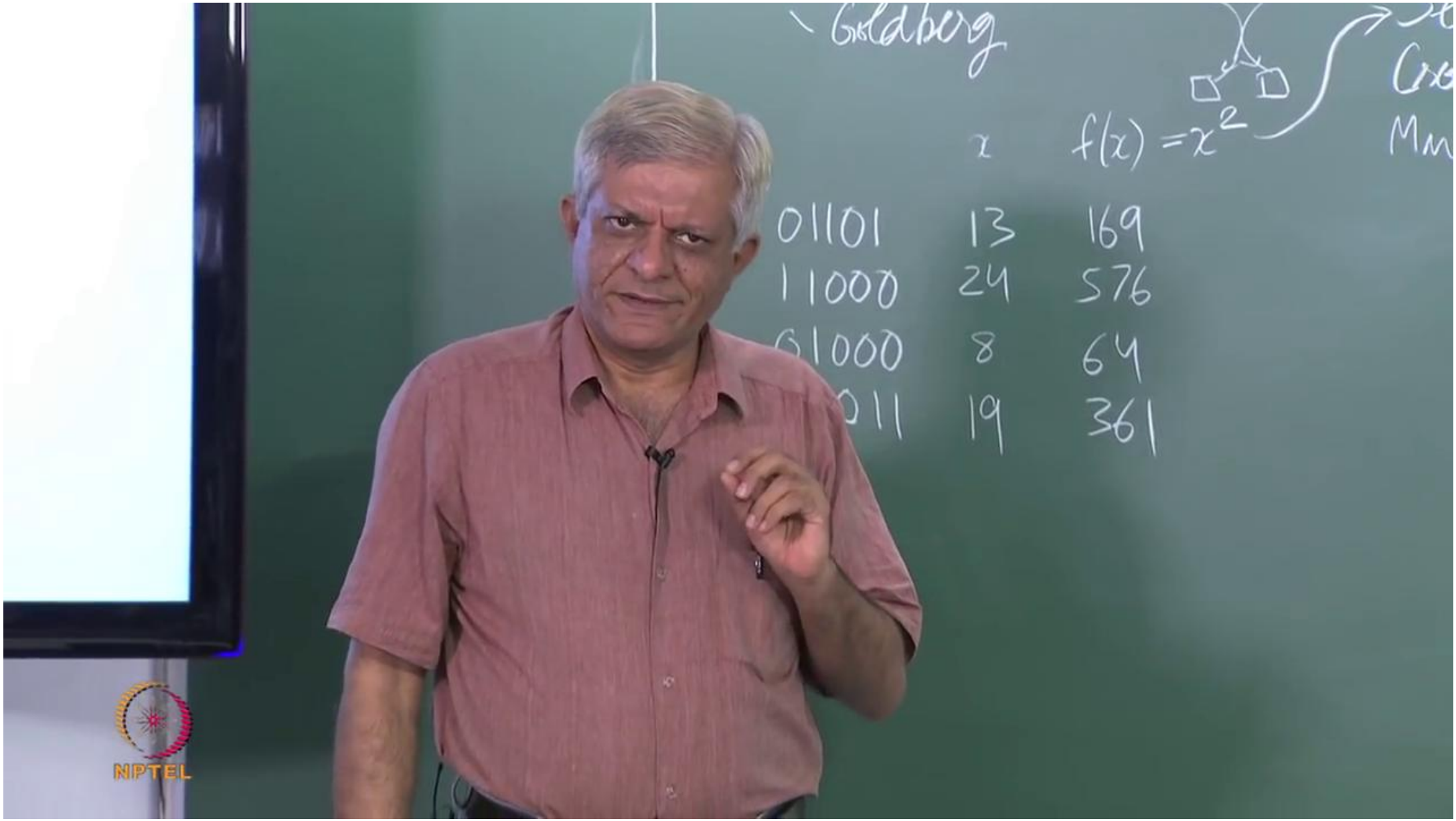
NPTEL

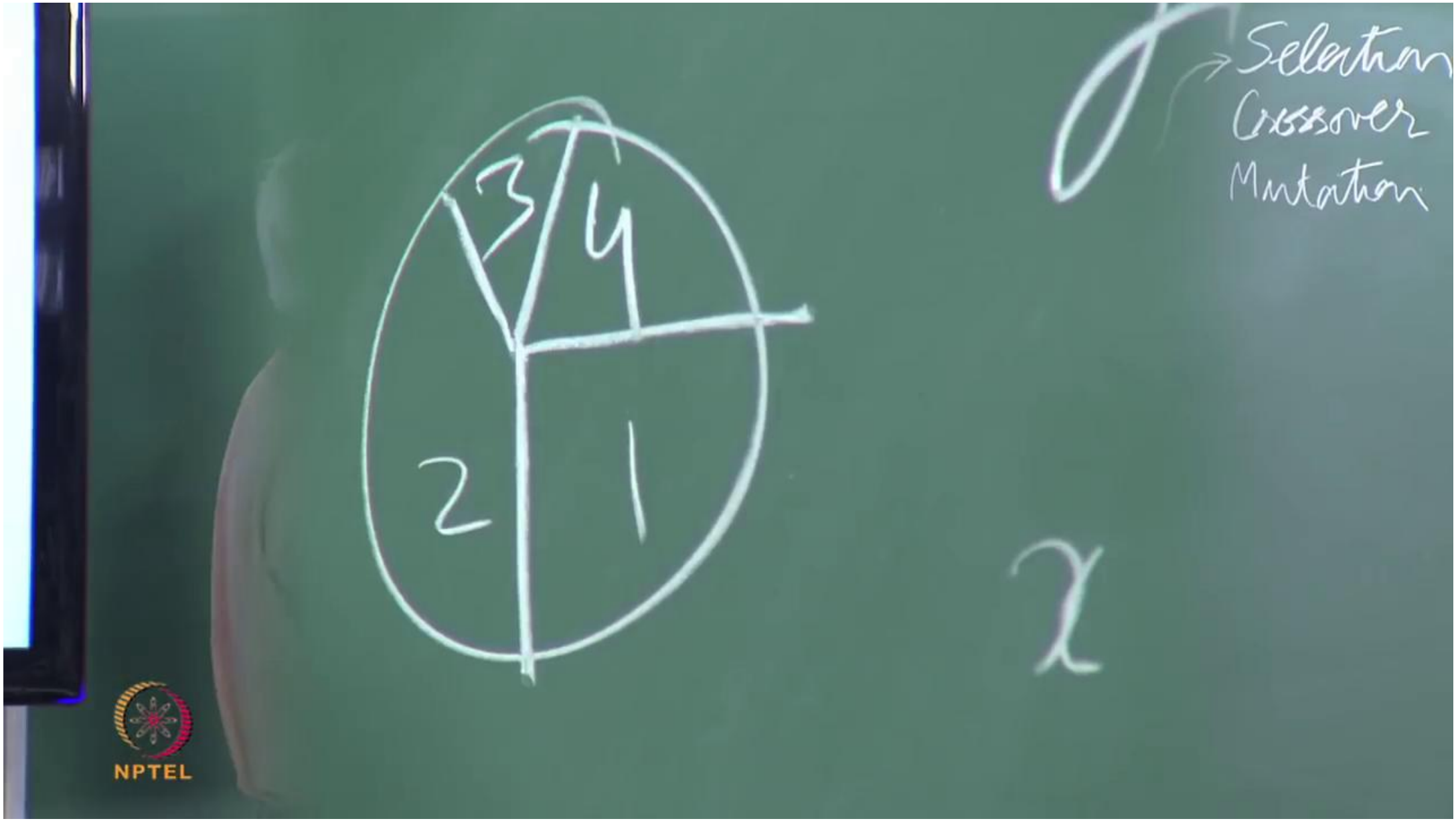

Holland
Goldberg
Selection
Crossover
Mutation
NPTEL

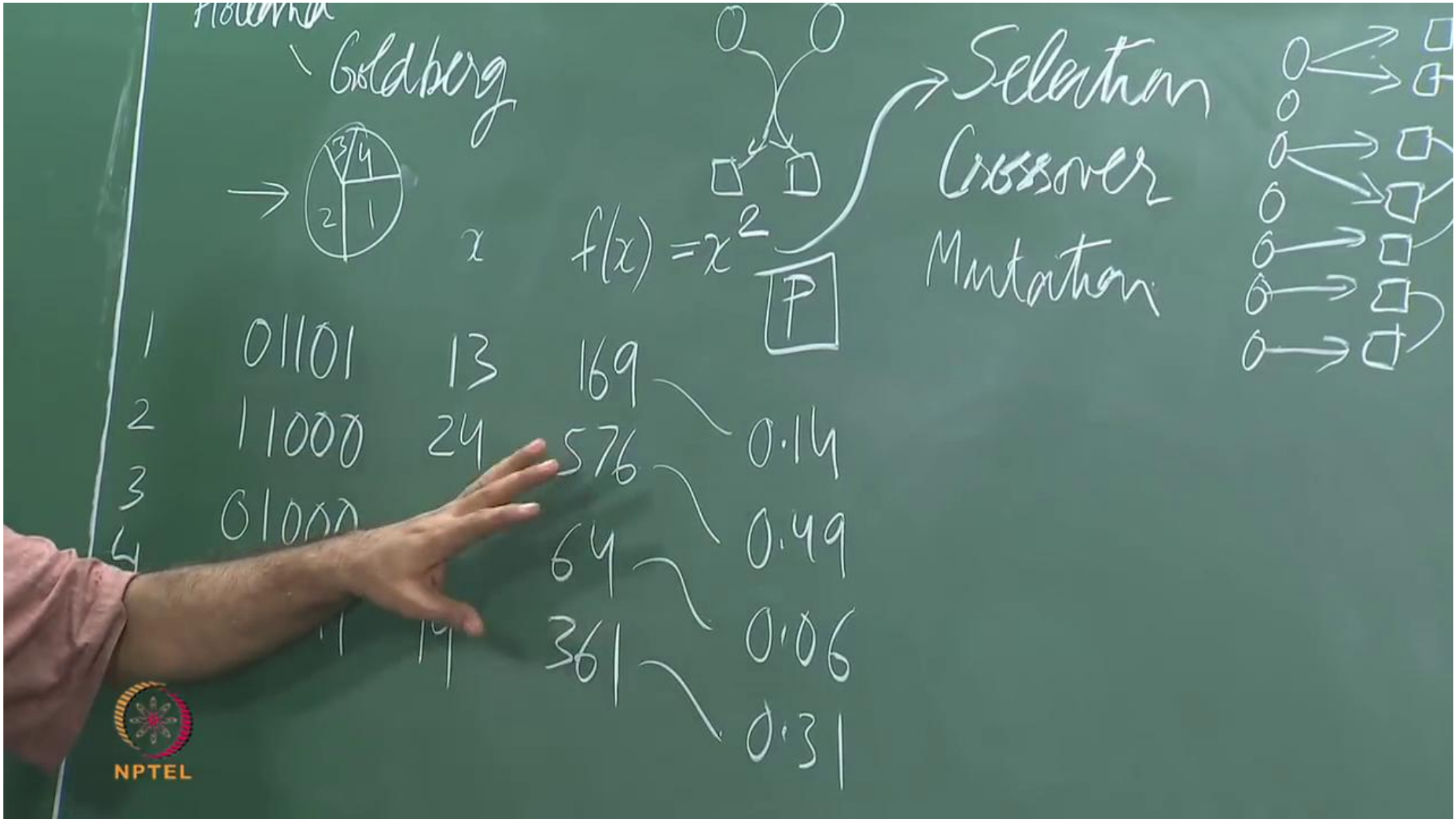

Holland
Goldberg
NPTEL


1 01101
2 11000
3 01000
4 10011
NPTEL

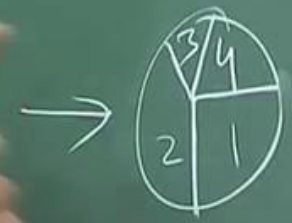

Goldberg
Select
Crossov
Mutati
x
$f(x) = x^2$
1101 13
1000 24
01000 8
10011 19
NPTEL


Goldberg
x
f(x) = x²
01101 13 169
11000 24 576
01000 8 64
011 19 361
NPTEL


3
4
2
1
Selection
Crossover
Mutation
NPTEL


Goldberg
3
4
2
1
x
f(x) = x²
P
Selection
Crossover
Mutation
1 01101 13 169 0.14
2 11000 24 576 0.49
3 01000 64 0.06
4 10011 19 361 0.31
NPTEL

Goldberg

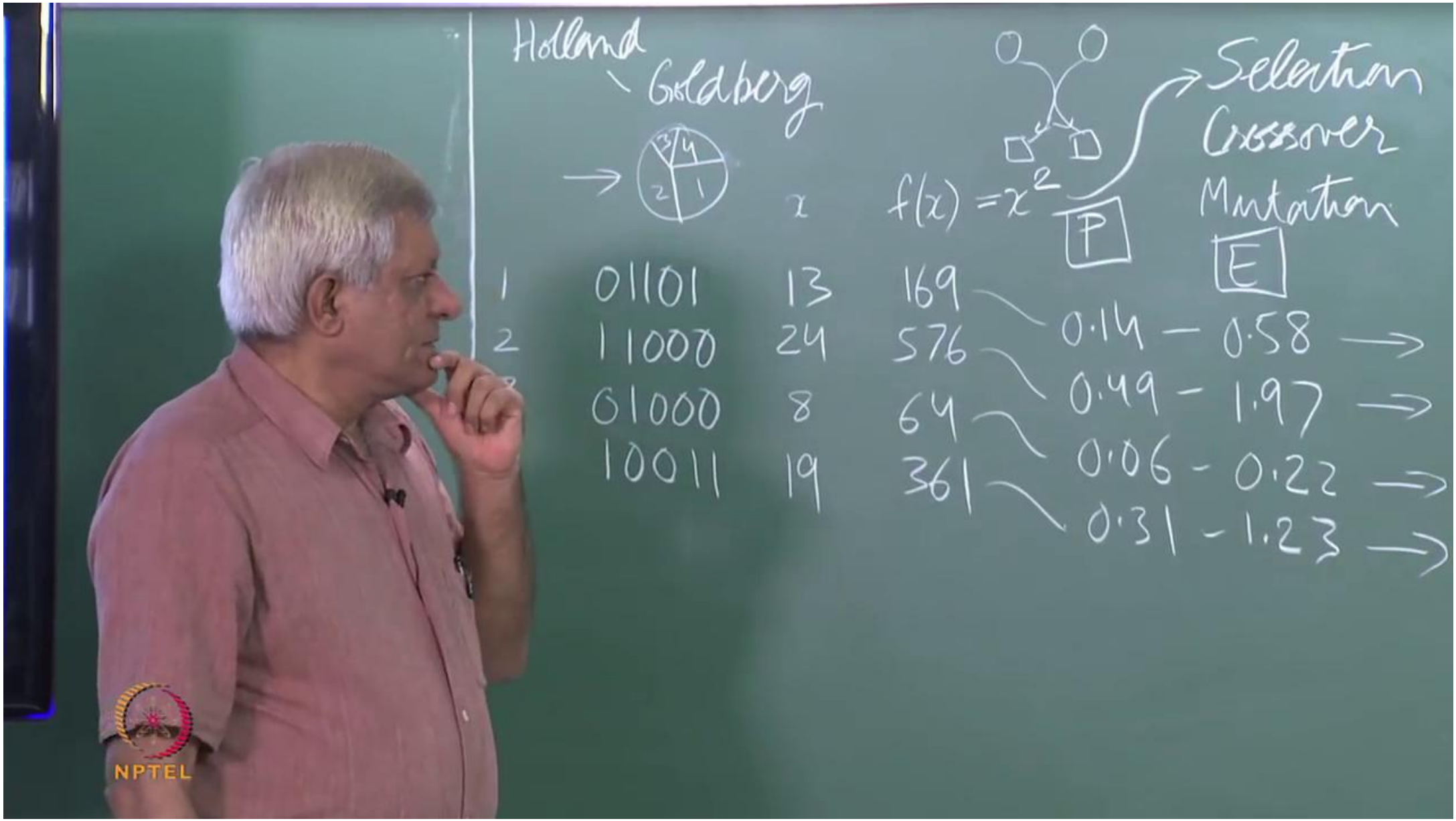


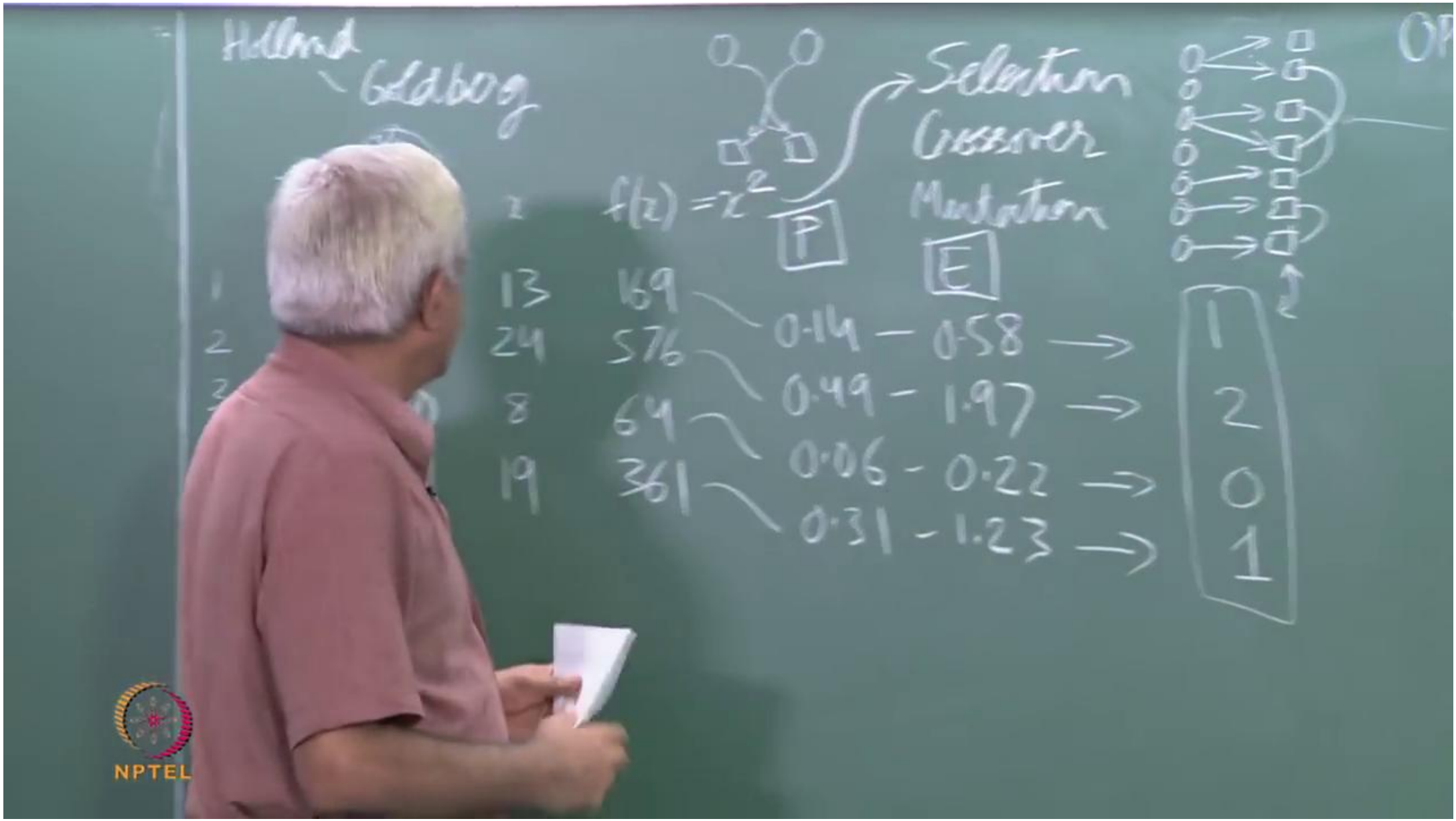
Selection
Crossover
Mutation

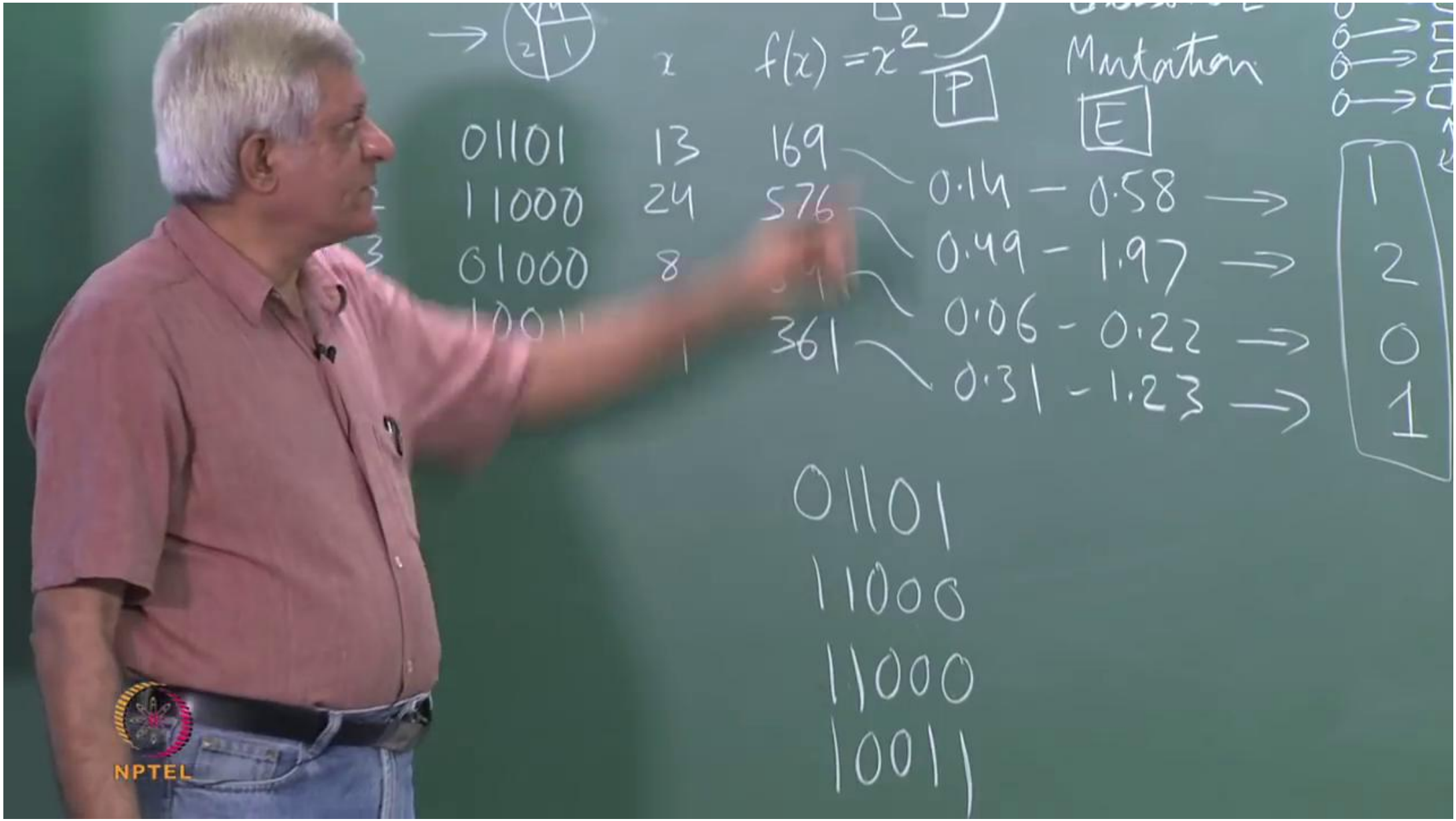
	x	$f(x) = x^2$	
01101	13	169	0.14
11000	24	576	0.49
01000	8	64	0.06
10011	19	361	0.31

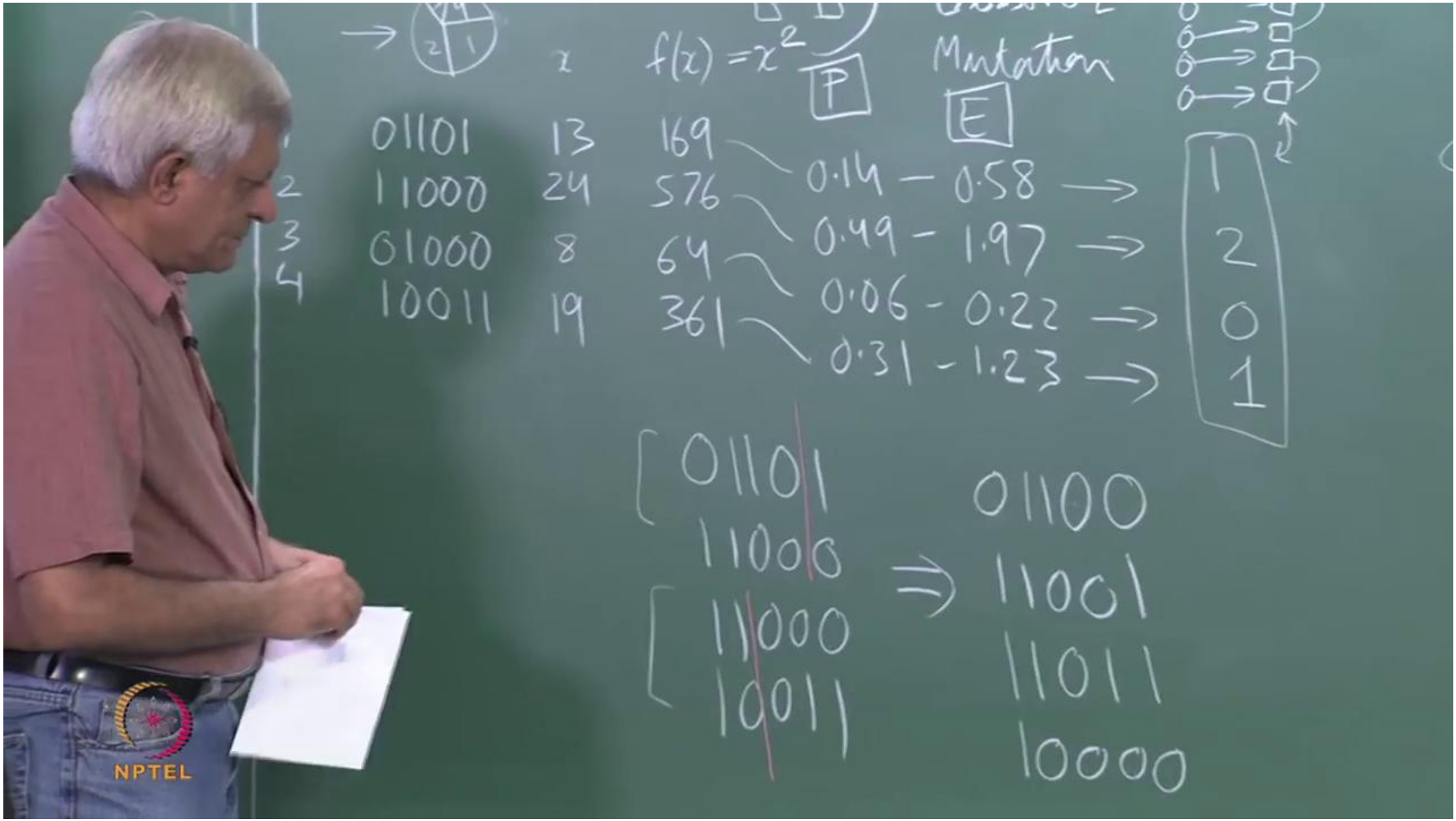
P

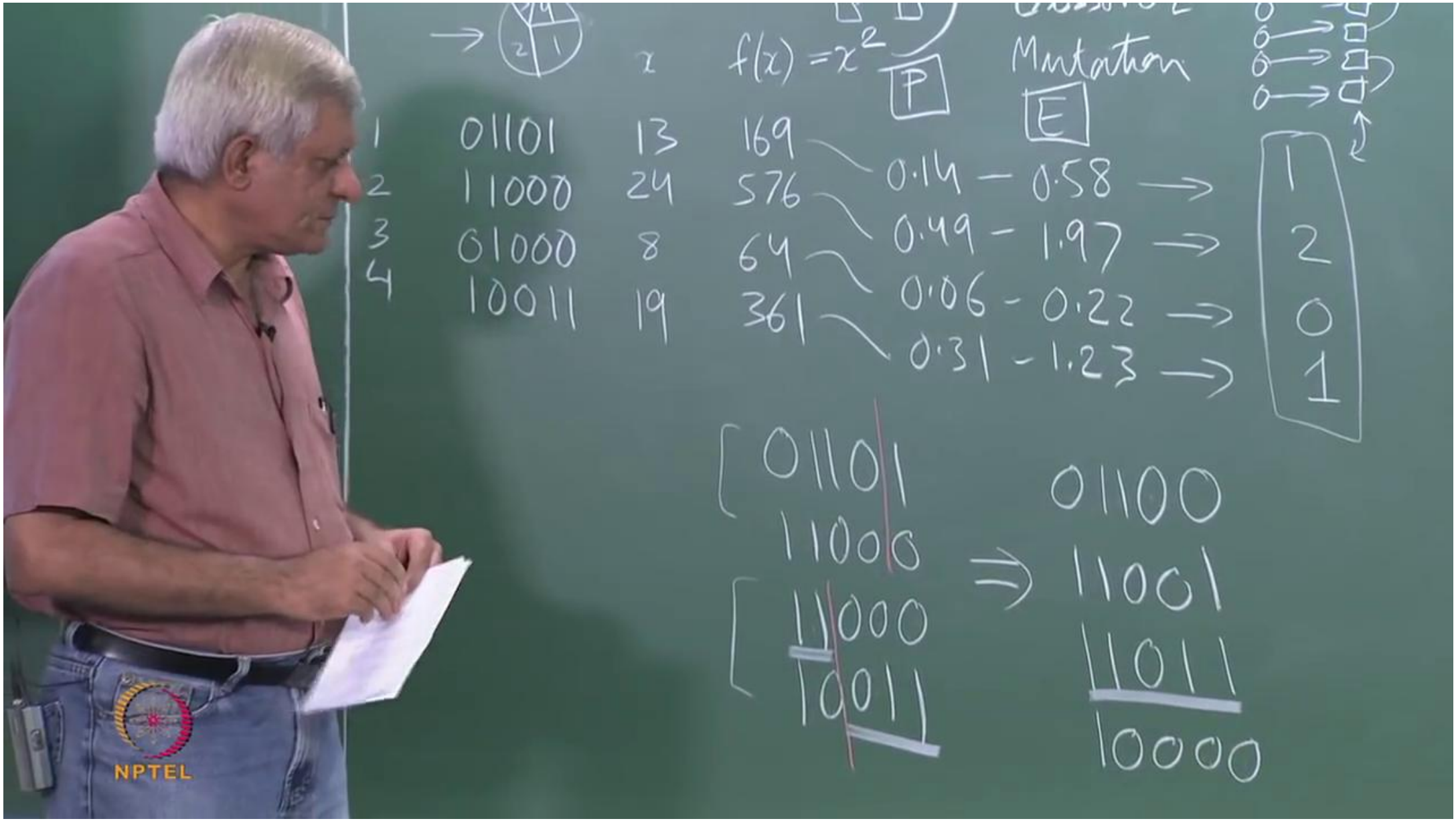
NPTEL

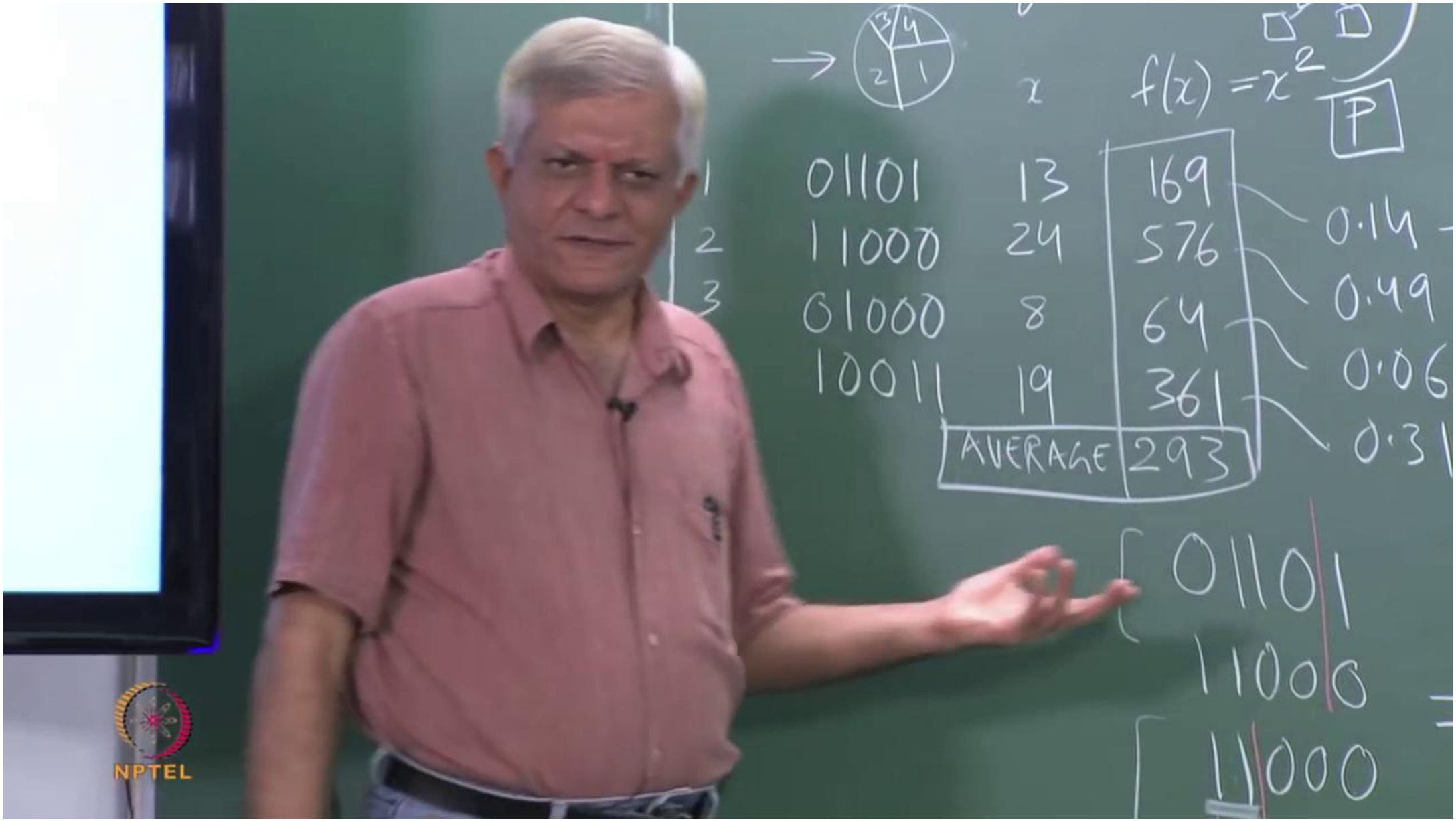

Holland
Goldberg
3/4
2
1
x
f(x) = x²
P
Selection
Crossover
Mutation
E
1 01101 13 169
2 11000 24 576
01000 8 64
10011 19 361
0.14 - 0.58
0.49 - 1.97
0.06 - 0.22
0.31 - 1.23
NPTEL

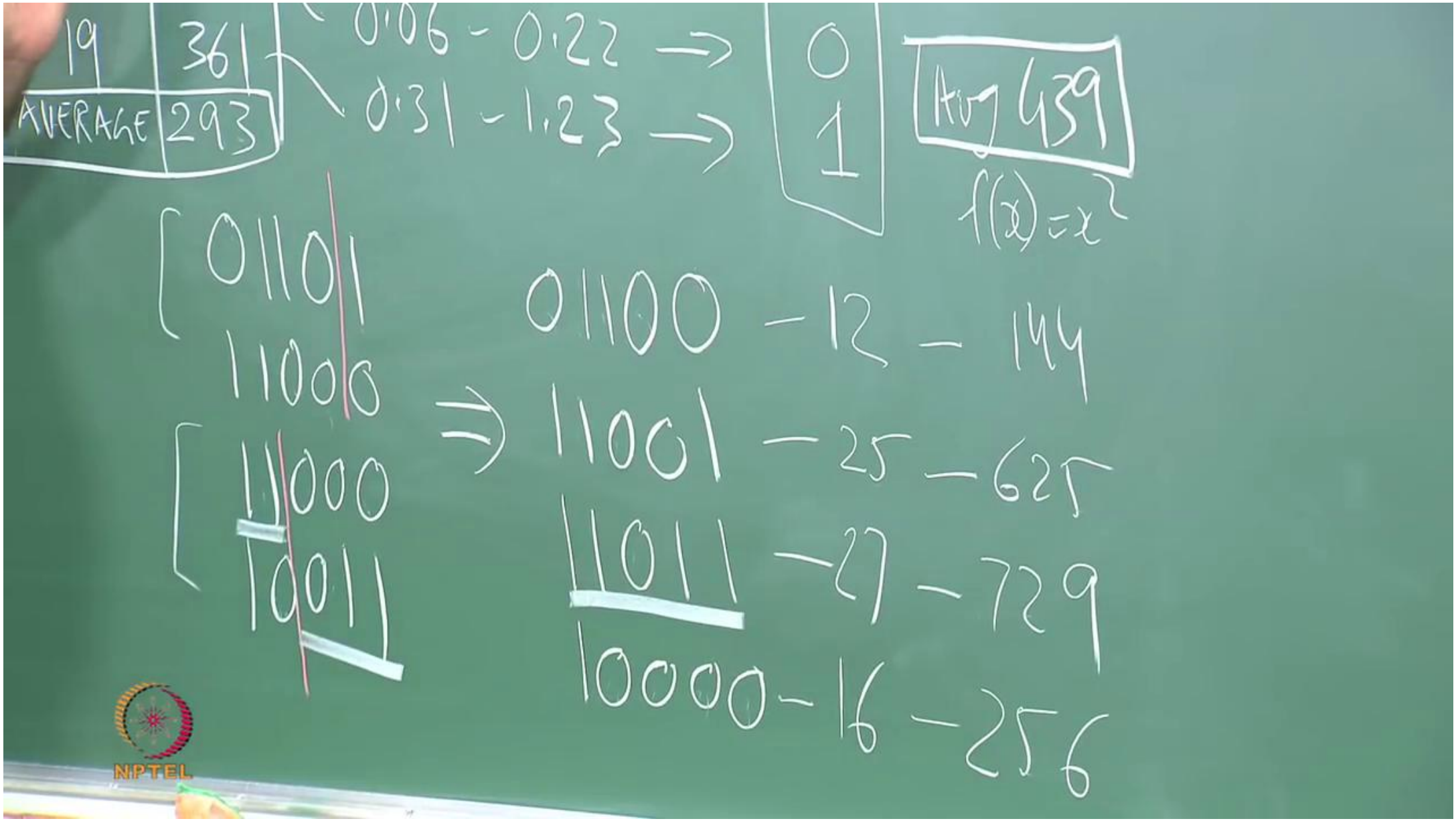

Holland
Goldberg
f(x) = x²
P
Selection
Crossover
Mutation
E
13
24
8
19
169
576
64
361
0.14 - 0.58
0.49 - 1.97
0.06 - 0.22
0.31 - 1.23
1
2
0
1

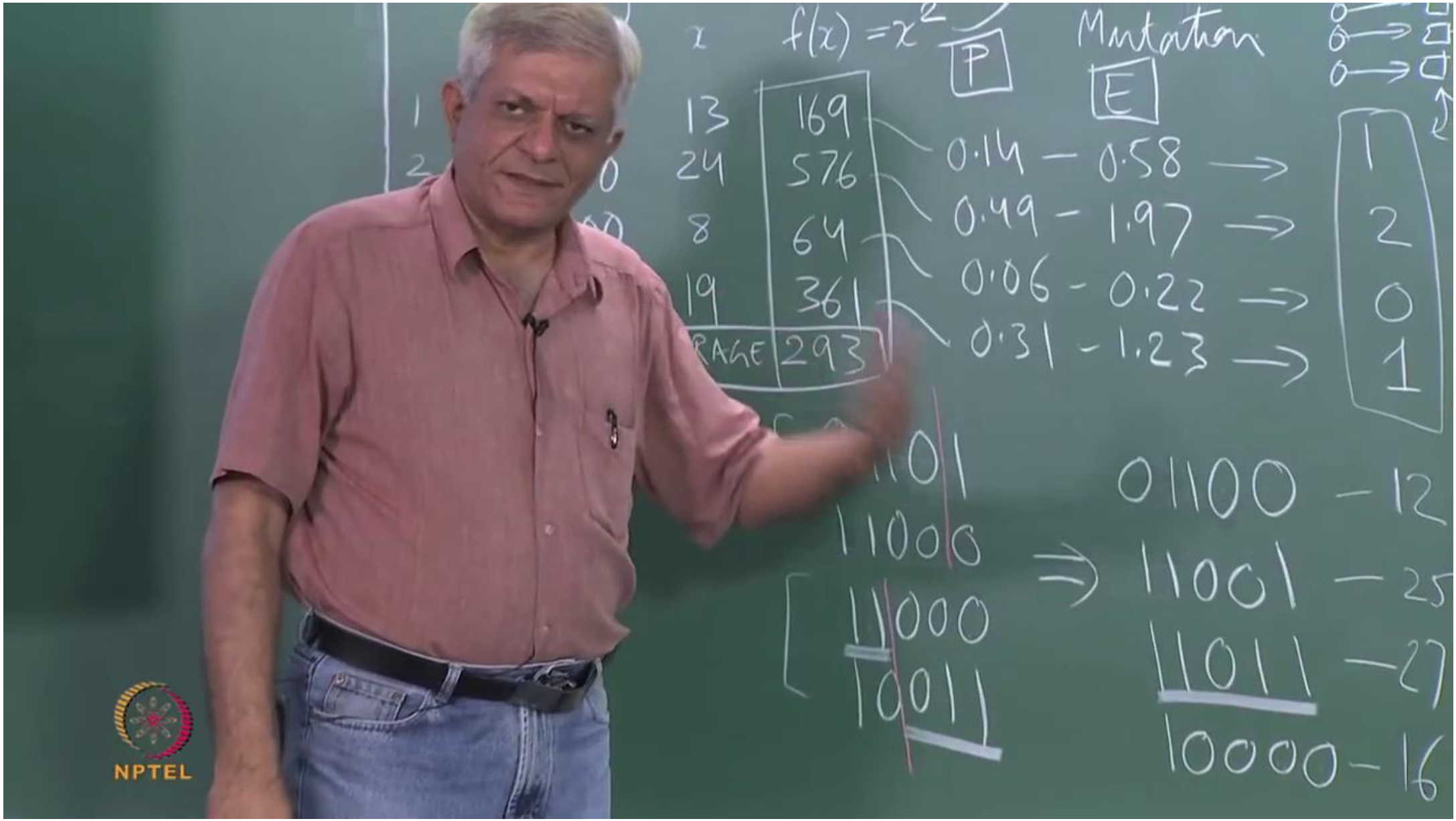

x
f(x) = x²
Mutation
P
E
01101 13 169
11000 24 576
01000 8
10011 361
0.14 - 0.58
0.49 - 1.97
0.06 - 0.22
0.31 - 1.23
1
2
0
1
01101
11000
11000
10011
NPTEL

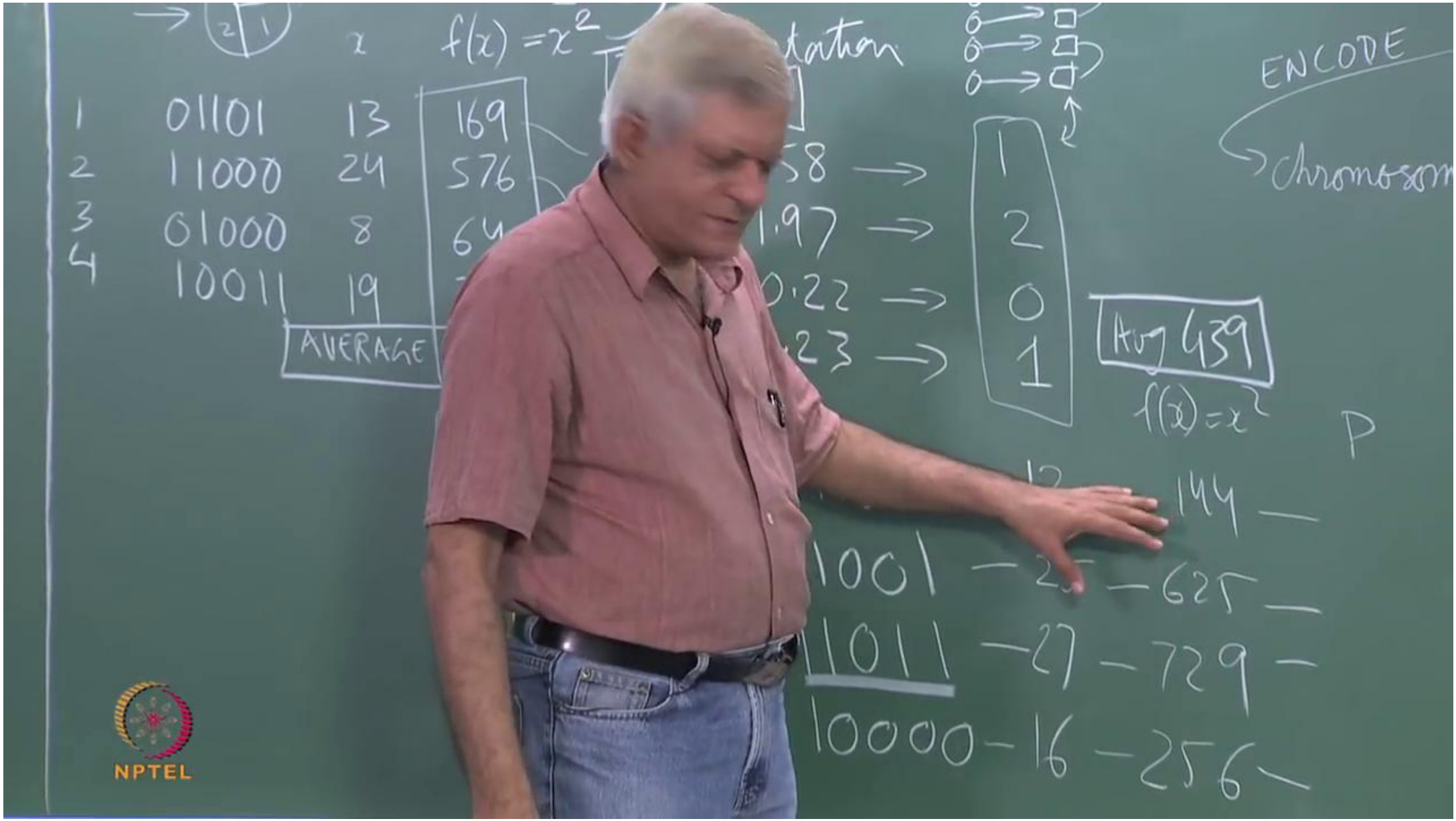

x
f(x) = x²
Mutation
P
E
1 01101 13 169
2 11000 24 576
3 01000 8 64
4 10011 19 361
0.14 - 0.58 → 1
0.49 - 1.97 → 2
0.06 - 0.22 → 0
0.31 - 1.23 → 1
01101
11000
11000
10011
⇒
01100
11001
11011
10000
NPTEL

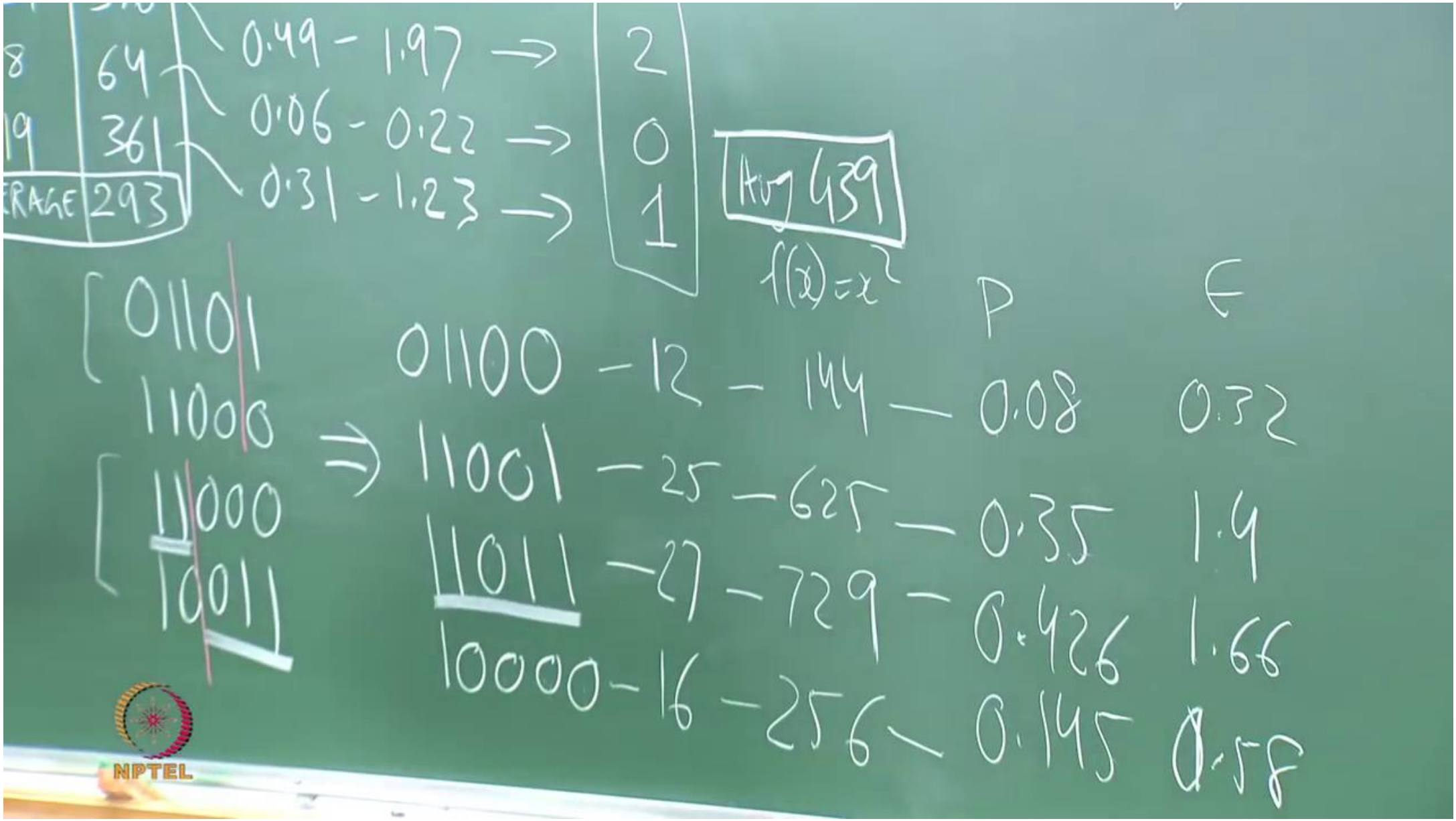

x
$f(x) = x^2$
Mutation
P
E
1 01101 13 169
2 11000 24 576
3 01000 8 64
4 10011 19 361
0.14 - 0.58
0.49 - 1.97
0.06 - 0.22
0.31 - 1.23
1
2
0
1
01101
11000
01100
11001
11000
10011
11011
10000
NPTEL

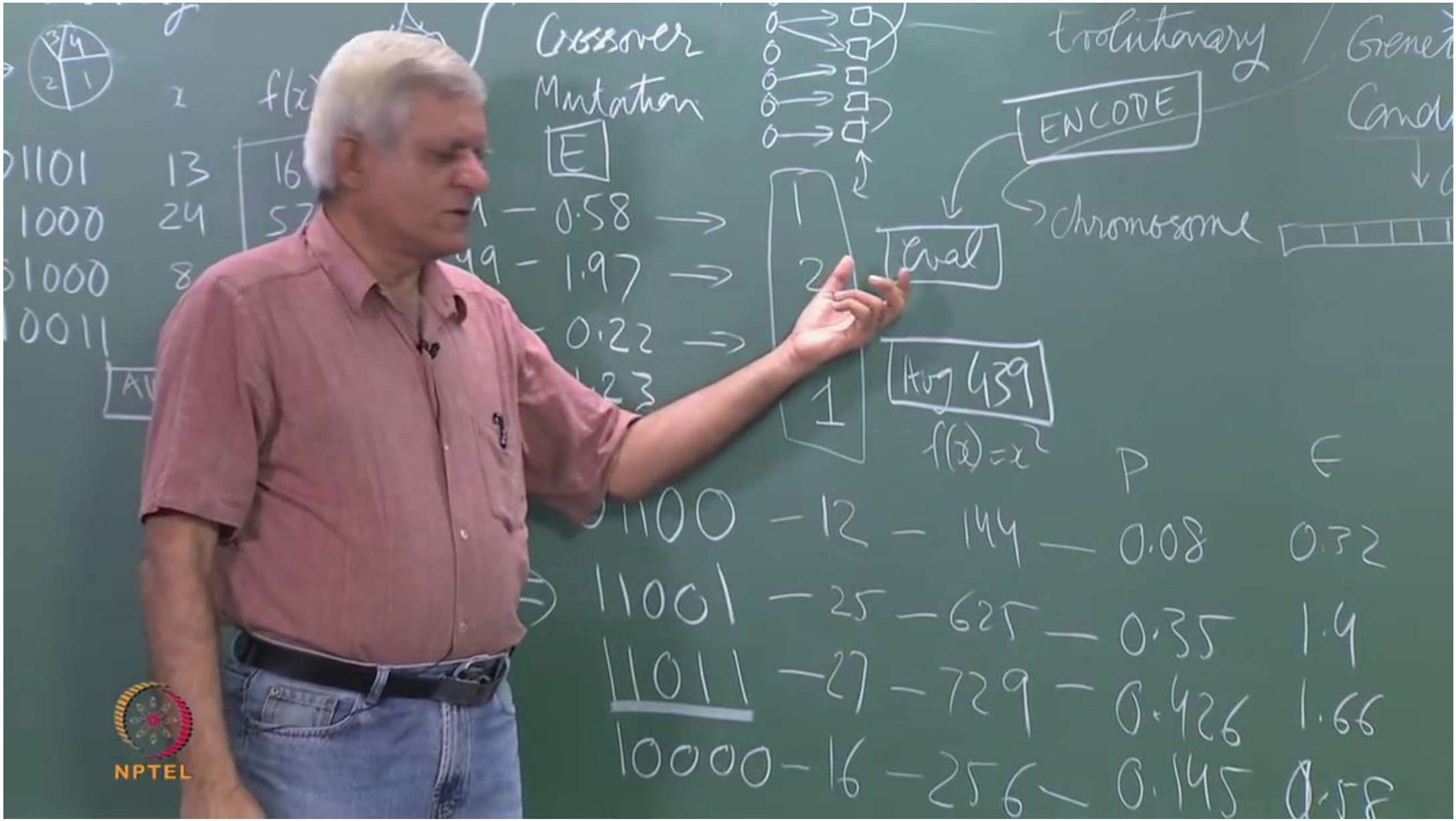

3 4
2 1
x
f(x) = x²
P
1 01101 13 169
2 11000 24 576
3 01000 8 64
10011 19 361
AVERAGE 293
0.14
0.49
0.06
0.31
01101
11000
11000

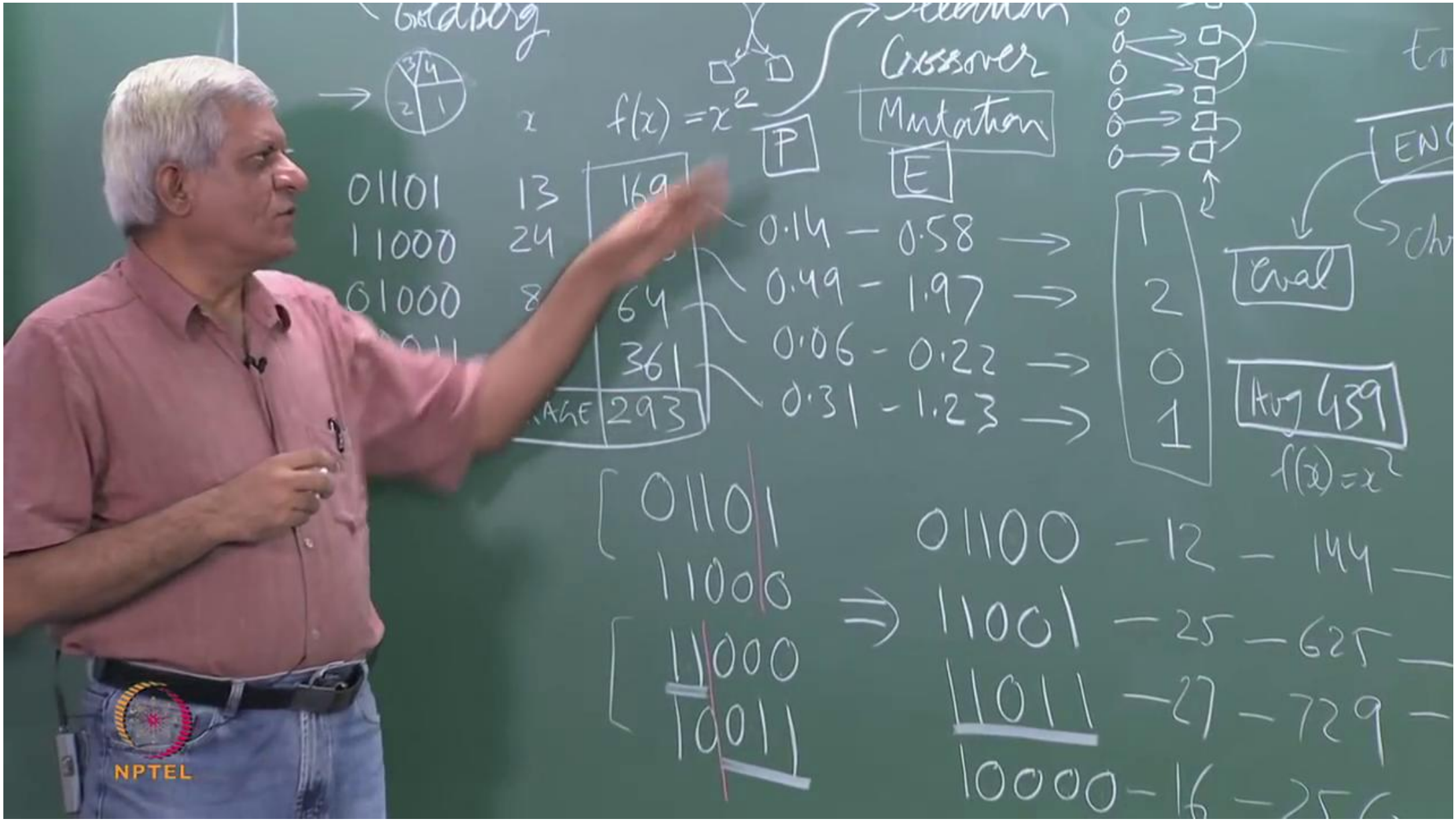

19 361
AVERAGE 293
0.06 - 0.22 ⇒ 0
0.31 - 1.23 ⇒ 1
Avg 439
f(x) = x²
01101
11000
11000
10011
⇒
01100 - 12 - 144
11001 - 25 - 625
11011 - 27 - 729
10000 - 16 - 256
NPTEL

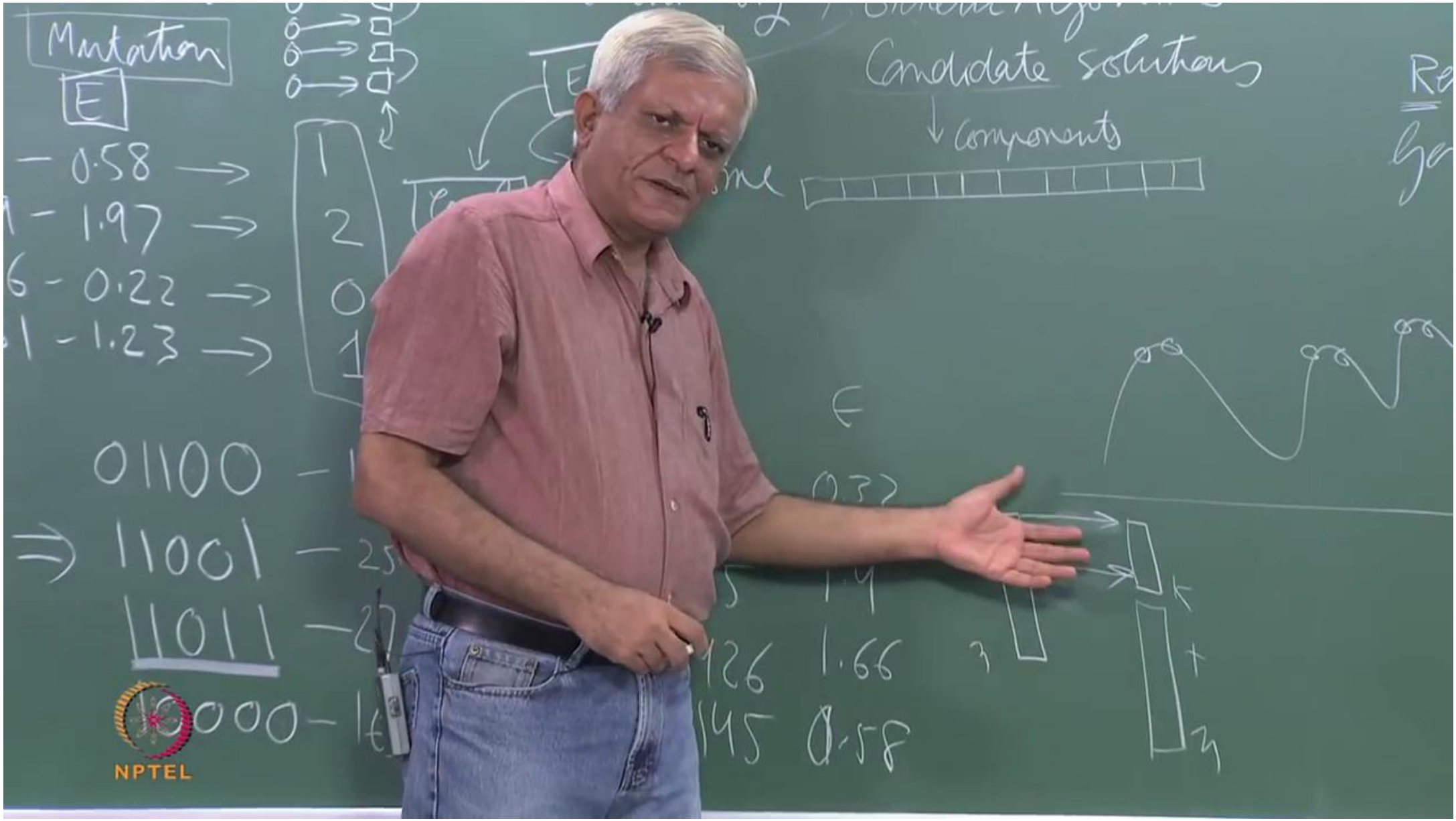

x
f(x) = x²
P
Mutation
E
13
169
24
576
8
64
19
361
RAGE
293
0.14 - 0.58
0.49 - 1.97
0.06 - 0.22
0.31 - 1.23
1
2
0
1
01100 - 12
11001 - 25
11011 - 27
10000 - 16
11000
10011
NPTEL

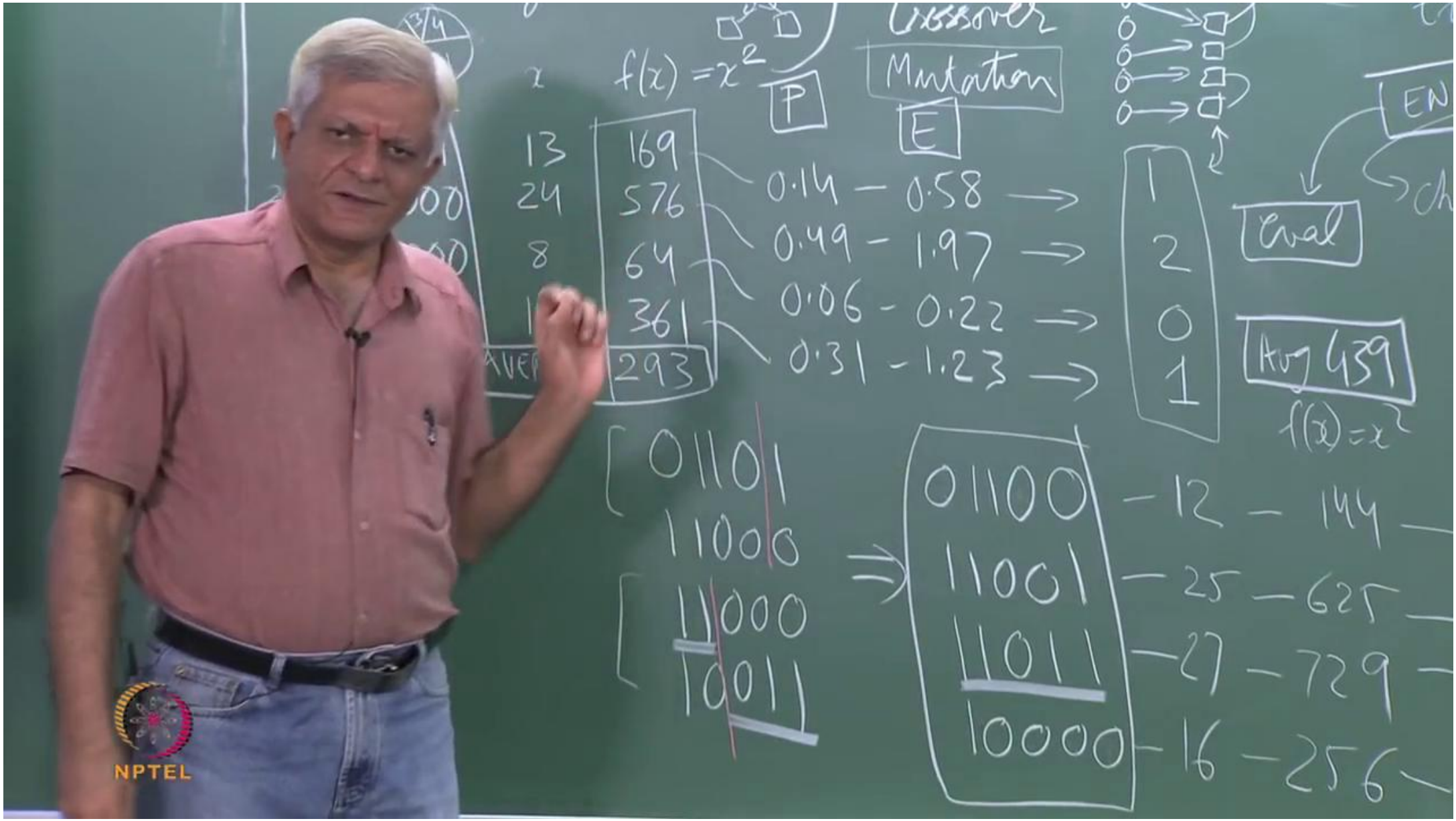

x
f(x) = x²
1 01101 13 169
2 11000 24 576
3 01000 8 64
4 10011 19
AVERAGE
ENCODE
Chromosom
Avg 459
f(x) = x²
1001 — 625 —
11011 — 729 —
10000 — 16 — 256 —
NPTEL


8 64
19 361
ERAGE 293
0.49 - 1.97 ⇒ 2
0.06 - 0.22 ⇒ 0
0.31 - 1.23 ⇒ 1
Avg 439
f(x) = x²
P
E
01101
11000
11000
10011
01100 - 12 - 144 - 0.08 0.32
⇒ 11001 - 25 - 625 - 0.35 1.4
11011 - 27 - 729 - 0.426 1.66
10000 - 16 - 256 - 0.145 0.58


Crossover
Mutation
E
ENCODE
Chromosome
Eval
Avg 459
$f(x) = x^2$
P
E
01100 - 12 - 144 - 0.08 0.32
11001 - 25 - 625 - 0.35 1.4
11011 - 27 - 729 - 0.426 1.66
10000 - 16 - 256 - 0.145 0.58
NPTEL


Goldberg
Crossover
Mutation
P
E
$f(x) = x^2$
x
01101 13 169
11000 24
01000 8
64
361
293
0.14 - 0.58
0.49 - 1.97
0.06 - 0.22
0.31 - 1.23
1
2
0
1
Eval
Avg 459
$f(x) = x^2$
01101
11000
11000
10011
01100 - 12 - 144
11001 - 25 - 625
11011 - 27 - 729
10000 - 16
NPTEL


Mutation
E
Candidate solutions
↓ Components
NPTEL


x
f(x) = x²
13
24
8
169
576
64
361
293
P
Crossover
Mutation
E
0.14 - 0.58
0.49 - 1.97
0.06 - 0.22
0.31 - 1.23
1
2
0
1
Eval
Avg 459
f(x) = x²
01101
11000
11000
10011
01100
11001
11011
10000
- 12 - 144
- 25 - 625
- 27 - 729
- 16 - 256
NPTEL