

Digital Communication Systems

EES 452

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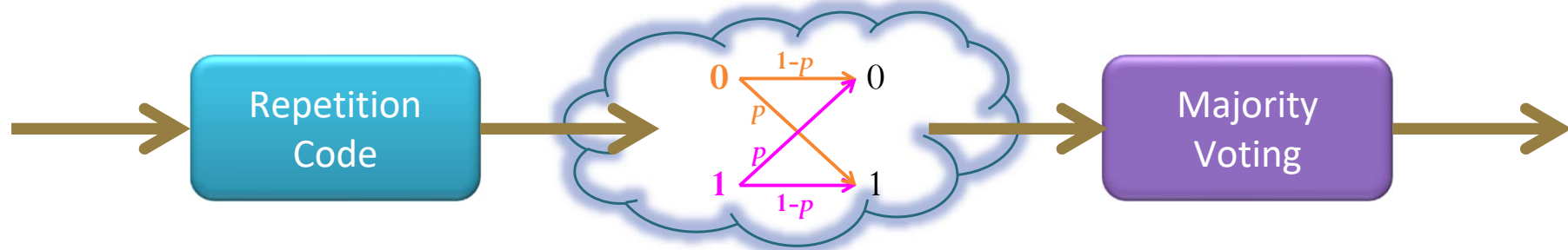
4. Mutual Information and Channel Capacity

4.2 Operational Channel Capacity

Example: Repetition Code

[Section 3.5]

BSC with $p = 0.2$

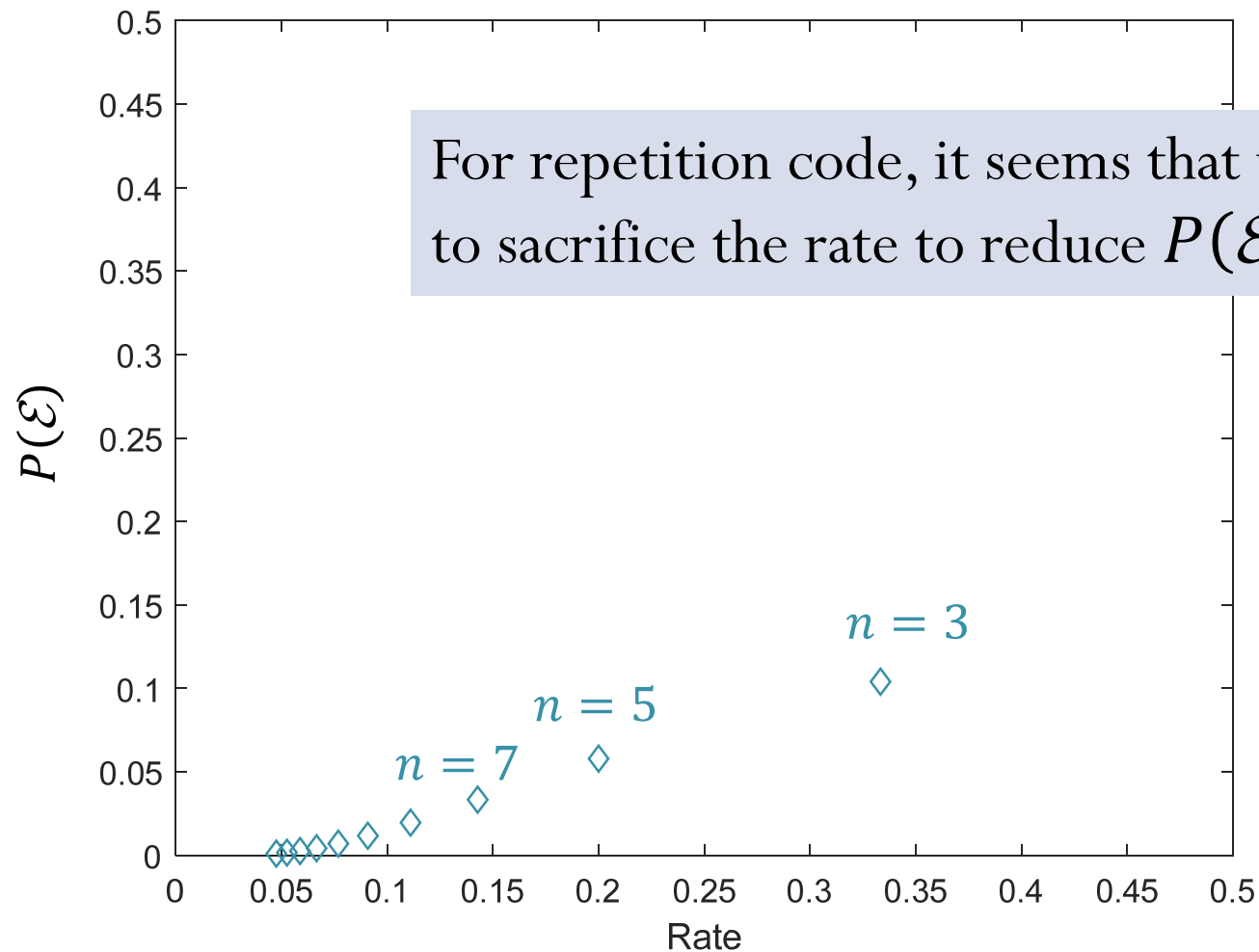


| n | $P(\mathcal{E})$ Probability that more than half of the bits are in error | Code Rate |
|-----|---|-------------------------------|
| 1 | $p = 0.2$ | $\frac{1}{1} = 1$ |
| 3 | $\binom{3}{2} p^2 (1-p) + \binom{3}{3} p^3 \approx 0.1040$ | $\frac{1}{3} \approx 0.33$ |
| 5 | $\binom{5}{3} p^3 (1-p)^2 + \binom{5}{4} p^4 (1-p)^1 + \binom{5}{5} p^5 \approx 0.0579$ | $\frac{1}{5} = 0.2$ |
| 7 | ≈ 0.0333 | $\frac{1}{7} \approx 0.1429$ |
| 9 | ≈ 0.0196 | $\frac{1}{9} \approx 0.1111$ |
| 11 | ≈ 0.0117 | $\frac{1}{11} \approx 0.0909$ |

Achievable Performance

BSC with $p = 0.2$

Repetition Code ($k = 1$)



Designing Channel Encoder

2^k rows

| \underline{s} | \underline{x} |
|-----------------|-----------------|
| 00 | ? ? ? ? ? |
| 01 | ? ? ? ? ? |
| 10 | ? ? ? ? ? |
| 11 | ? ? ? ? ? |

n columns

Each “?” can be 0 or 1.

So, there are

$$2^{(n2^k)} = 1,048,576 \text{ for } n = 5, k = 2$$

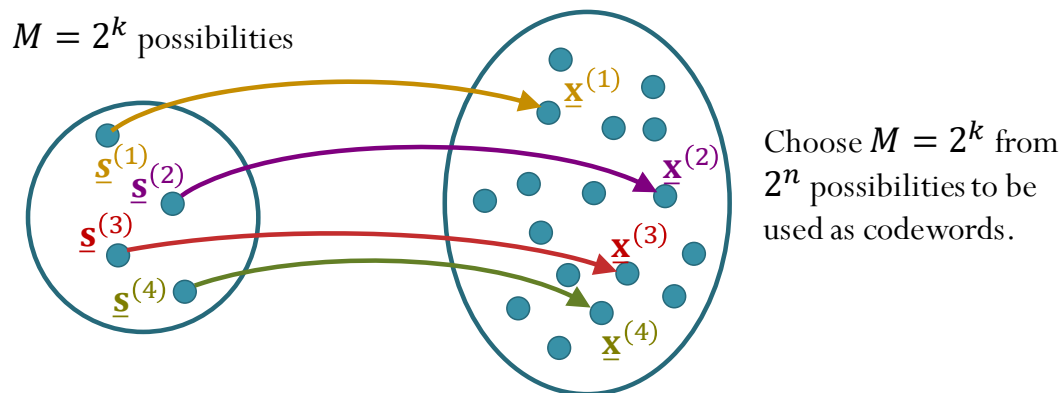
possibilities.

But we don't want to use the same codeword to represent two different info blocks.

So, actually, we need to consider

$$\binom{2^n}{2^k} = 35,960 \text{ for } n = 5, k = 2$$

possibilities.



MATLAB

```
close all; clear all;

% EES315 2020 Example 6.58
% EES452 2020 Examples 3.62, 3.67
C = [0 0 0 0 0; 1 1 1 1 1]; % repetition code

p = (1/100);
PE_minDist(C,p)
```

Code C is defined by putting all its (valid) codewords as its rows. For repetition code, there are two codewords: 00..0 and 11..1.

Crossover probability of the binary symmetric channel.

```
>> PE_minDist_demo1

ans =
    9.8506e-06
```



MATLAB

PE_minDist.m

[Section 3.5]

```
function PE = PE_minDist(C,p)
% Function PE_minDist computes the error probability P(E) when code C
% is used for transmission over BSC with crossover probability p.
% Code C is defined by putting all its (valid) codewords as its rows.
M = size(C,1); % the number of (valid) codewords
k = log2(M);
n = size(C,2);

% Generate all possible n-bit received vectors
Y = dec2bin(0:2^n-1)-'0';

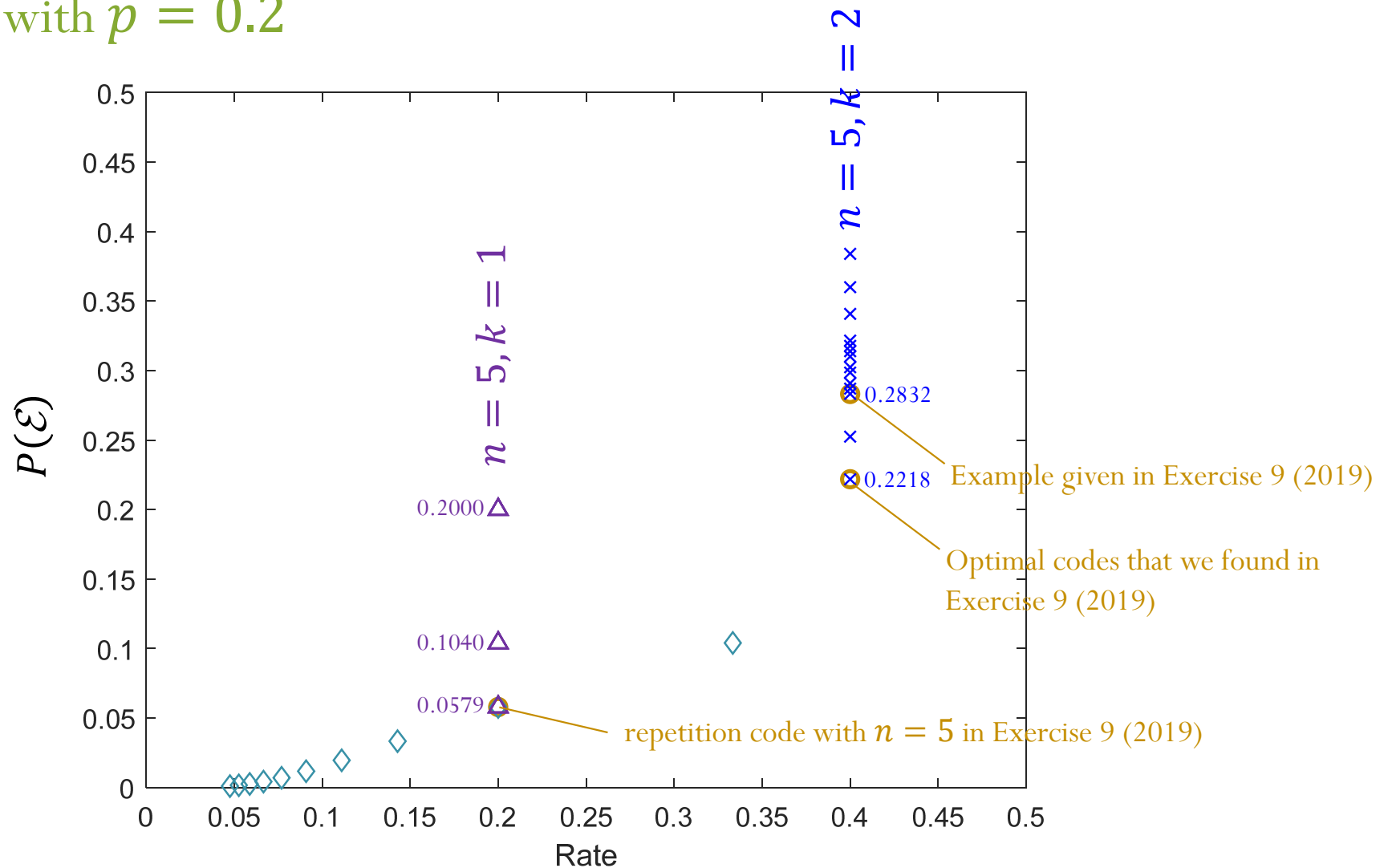
% Normally, we need to construct an extended Q matrix. However, because
% each conditional probability in there is a decreasing function of the
% (Hamming) distance, we can work with the distances instead of the
% conditional probability. In particular, instead of selecting the max in
% each column of the Q matrix, we consider min distance in each column.
dminy = zeros(1,2^n); % preallocation
for j = 1:(2^n)
    % for each received vector y,
    y = Y(j,:);
    % find the minimum distance
    % (the distance from y to the closest codeword)
    d = sum(mod(bsxfun(@plus,y,C),2),2);
    dminy(j) = min(d);
end

% From the distances, calculate the conditional probabilities.
% Note that we compute only the values that are to be selected (instead of
% calculating the whole Q first).
n1 = dminy; n0 = n-dminy;
Qmax = (p.^n1).*((1-p).^n0);
% Scale the conditional probabilities by the input probabilities and add
% the values. Note that we assume equally likely input.
PC = sum((1/M)*Qmax);
PE = 1-PC;
end
```



Achievable Performance

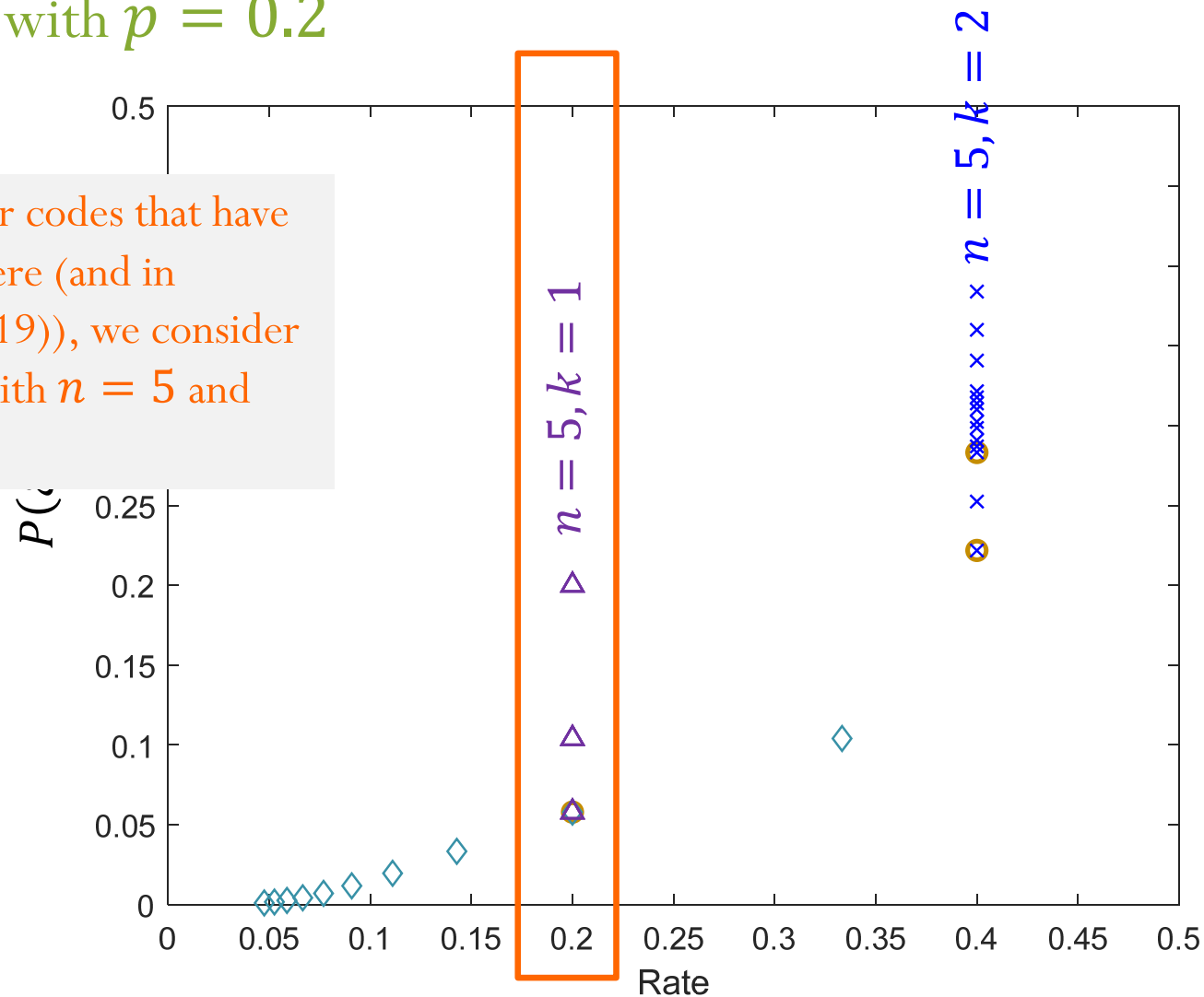
BSC with $p = 0.2$



Achievable Performance

BSC with $p = 0.2$

There are other codes that have rate = 0.2. Here (and in Exercise 9 (2019)), we consider all the codes with $n = 5$ and $k = 1$.

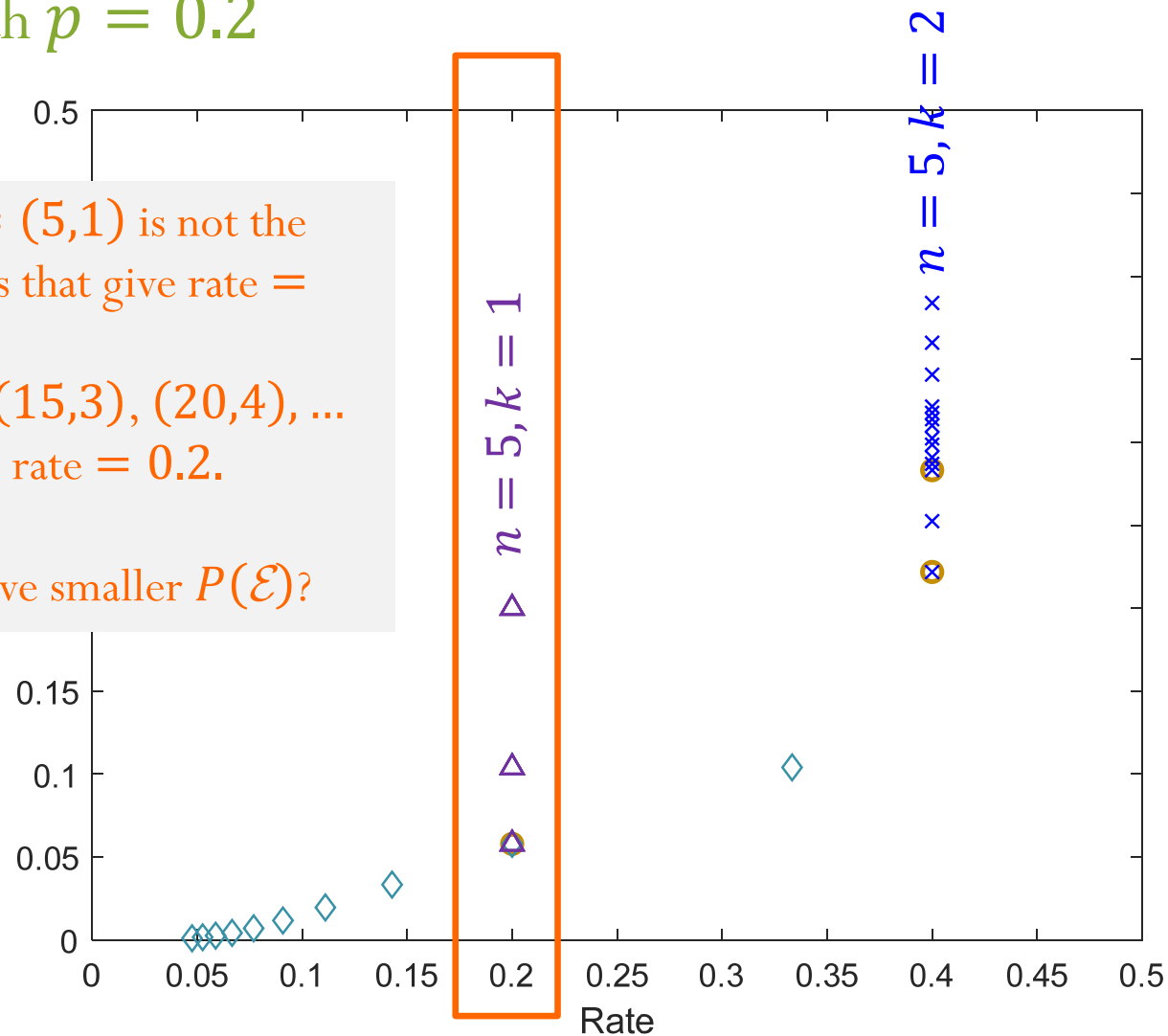


Achievable Performance

BSC with $p = 0.2$

Note that $(n, k) = (5, 1)$ is not the only family of codes that give rate = 0.2.
 $(n, k) = (10, 2), (15, 3), (20, 4), \dots$ also corresponds to rate = 0.2.

Will these codes have smaller $P(\mathcal{E})$?

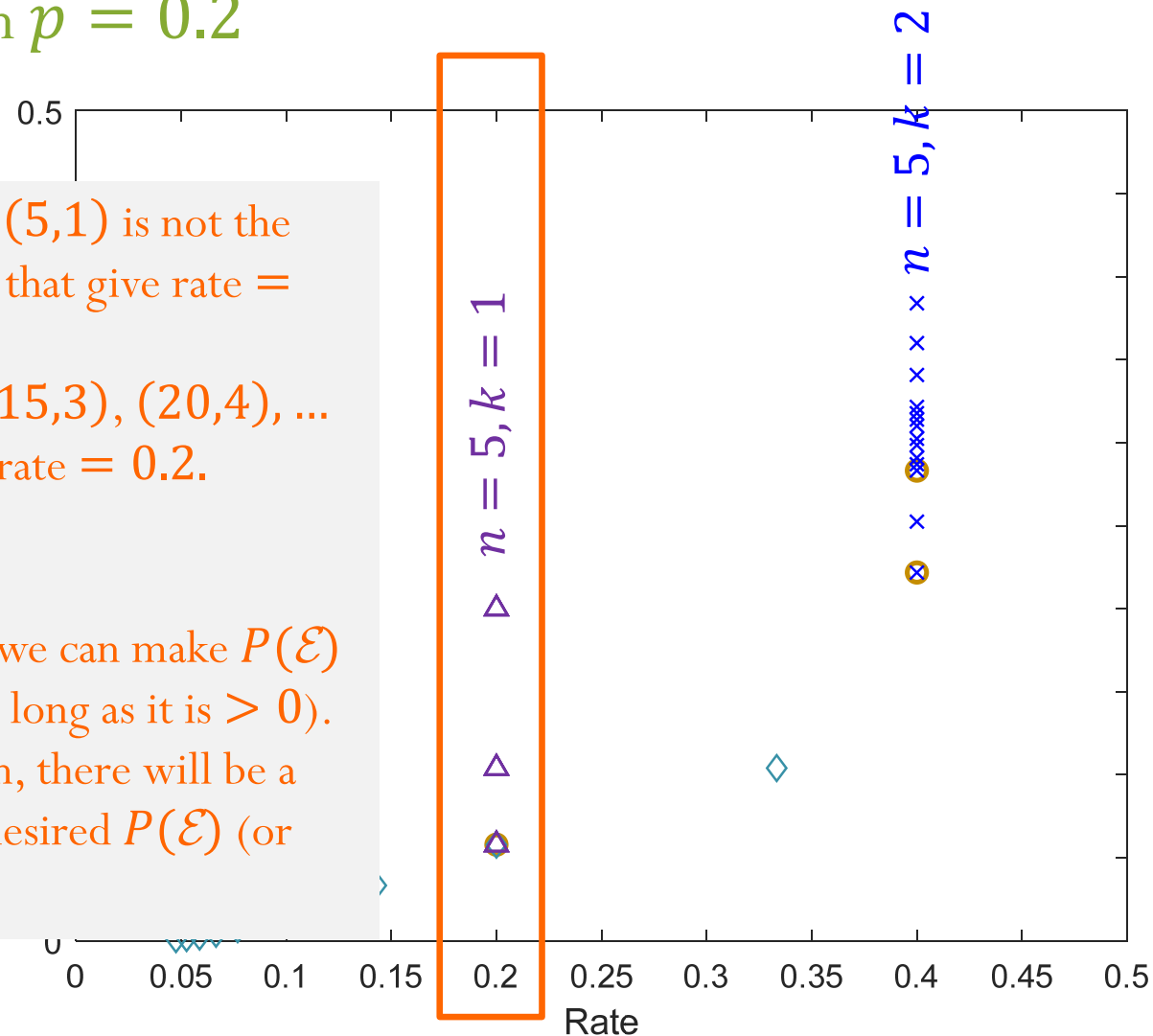


Achievable Performance

BSC with $p = 0.2$

Note that $(n, k) = (5, 1)$ is not the only family of codes that give rate = 0.2.
 $(n, k) = (10, 2), (15, 3), (20, 4), \dots$ also corresponds to rate = 0.2.

At rate = 0.2,
Shannon found that we can make $P(\mathcal{E})$ as small we want (as long as it is > 0).
With n large enough, there will be a code that gives the desired $P(\mathcal{E})$ (or smaller).

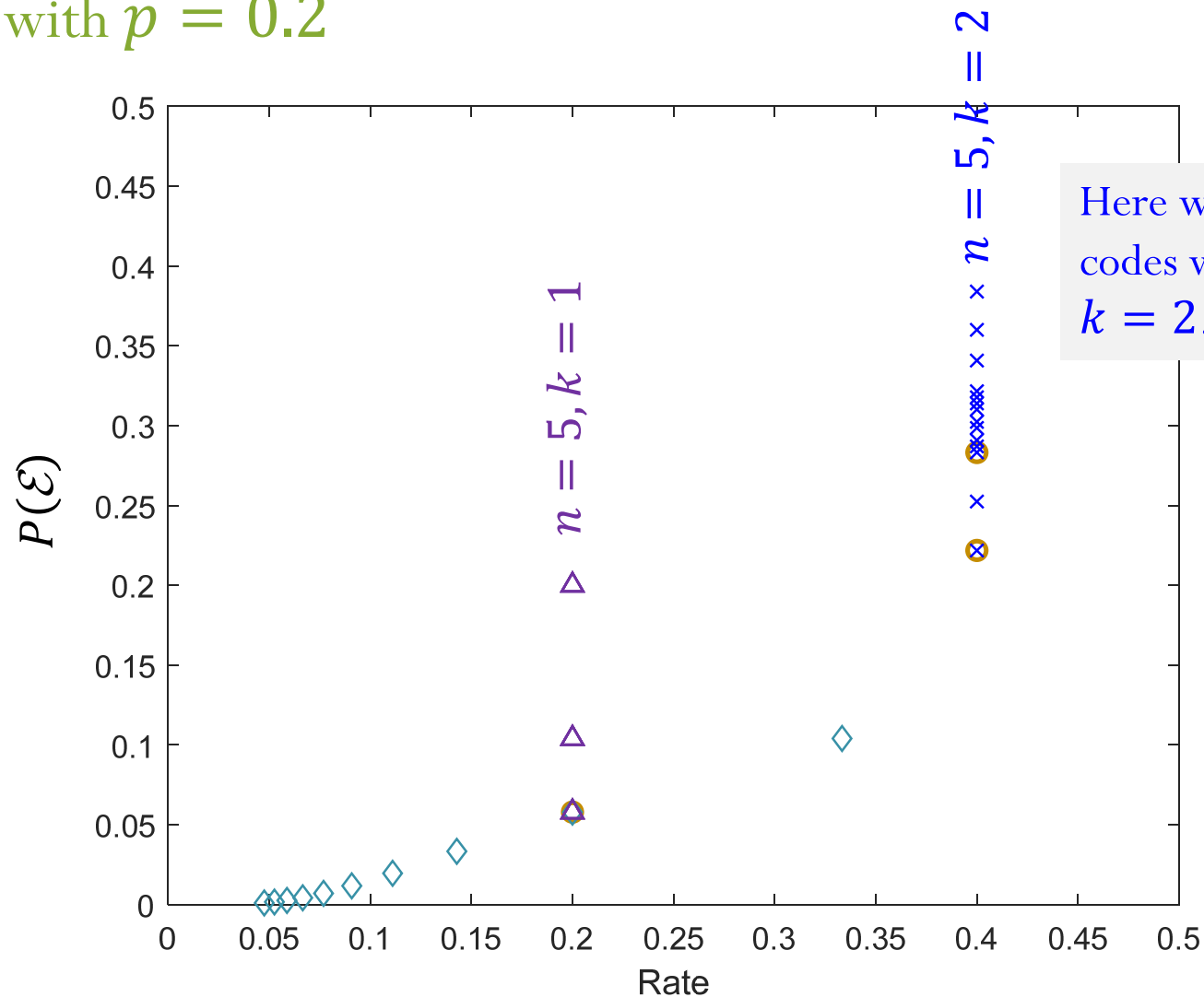


Reliable communication

- **Reliable communication** (at a particular rate) means arbitrarily small error probability can be achieved (at that rate).
- In our example, Shannon showed that reliable communication is achievable at rate = 0.2.
- Turn out that reliable communication is not achievable at rate = 0.4.

Achievable Performance

BSC with $p = 0.2$

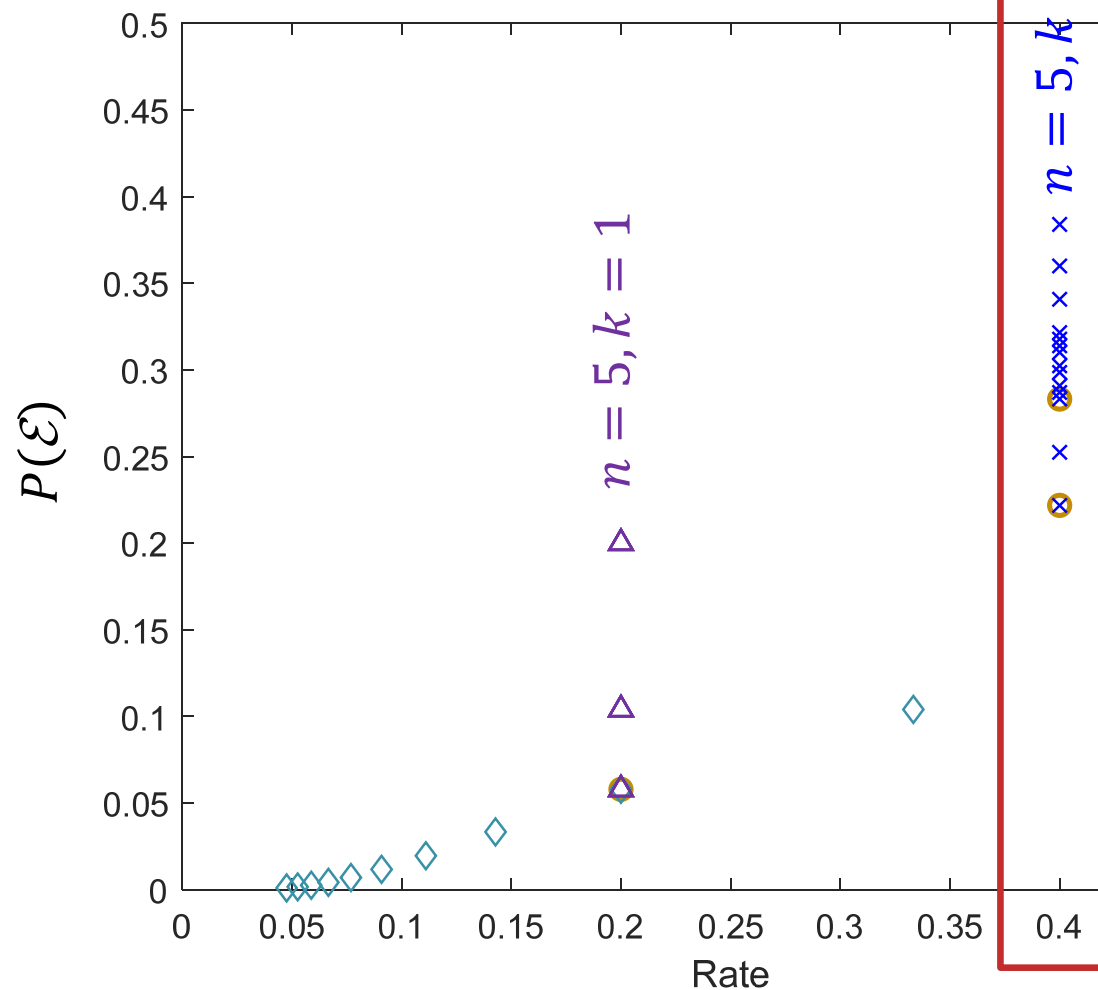


Here we consider all the codes with $n = 5$ and $k = 2$.



Achievable Performance

BSC with $p = 0.2$



Note that $(n, k) = (5, 2)$ is not the only family of codes that give rate = 0.4. $(n, k) = (10, 4), (15, 6), (20, 8), \dots$ also corresponds to rate = 0.4.

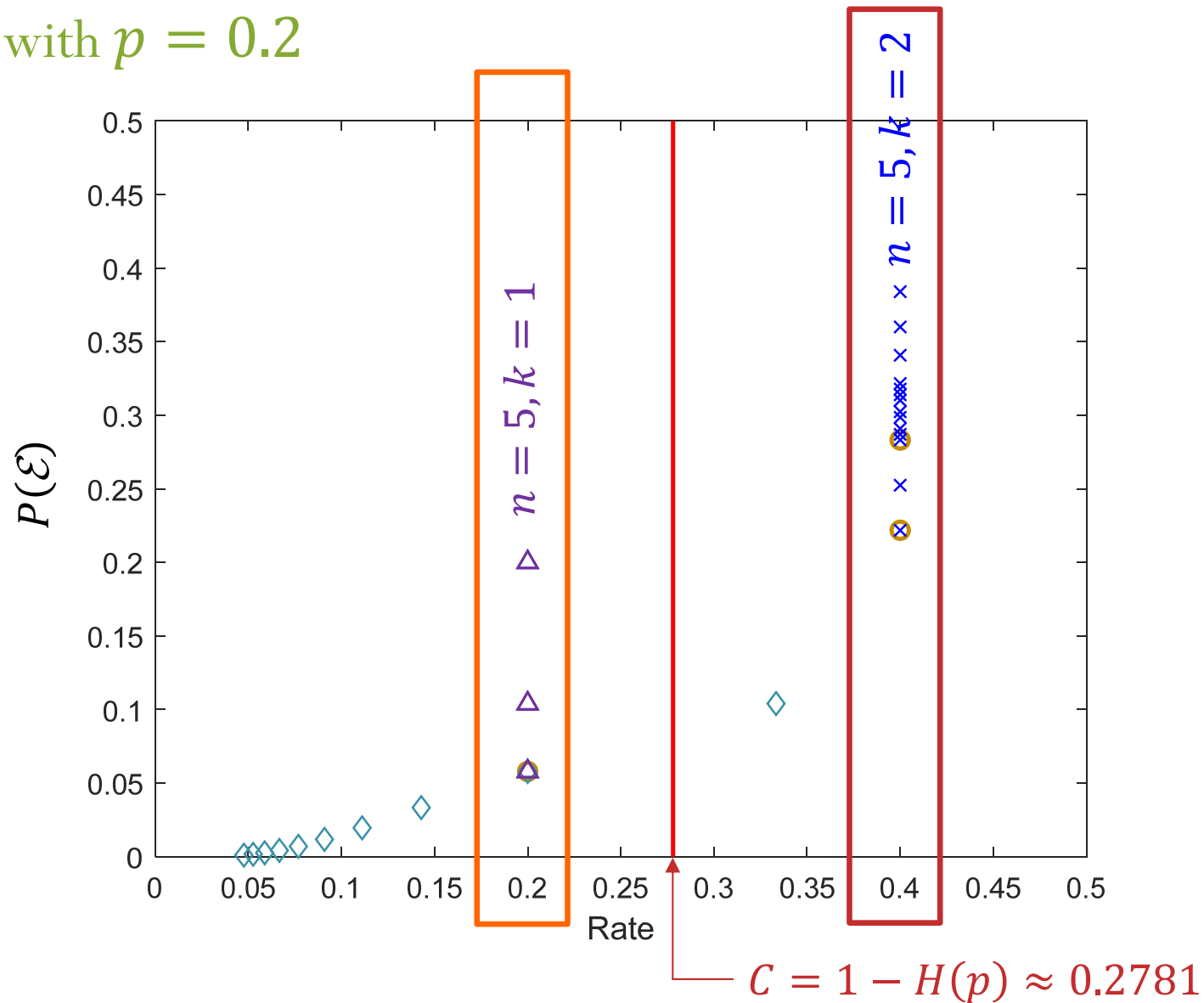
At rate = 0.4, Shannon found that we **cannot** make $P(\mathcal{E})$ as small as we want; even when we use large n .

So, how can we determine which rate can have arbitrarily small $P(\mathcal{E})$?



Achievable Performance

BSC with $p = 0.2$



Channel Capacity

[Section 4.2]

“**Operational**”: max rate at which **reliable** communication is possible

Arbitrarily small error probability can be achieved.

Channel Capacity

“**Information**”: $\max_{\underline{p}} I(X; Y)$ [bpcu]
[Section 4.3]

Shannon [1948] showed that these two quantities are actually the same.

