

# Lecture 20. Quantum Convolutional Codes

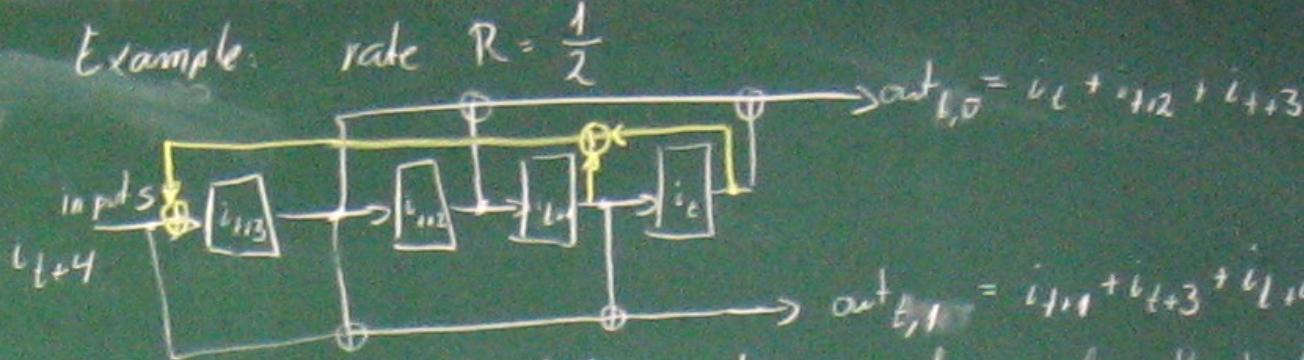
## 20.1 Motivation

- Block codes encode a fixed number of qubits, good codes have a large blocklength  $n$   
good code:  $\lim_{n \rightarrow \infty} \frac{d}{n} > 0, \lim_{n \rightarrow \infty} \frac{k}{n} > 0$   
 $\hookrightarrow$  generators of the stabilizer group span many qubits ('non-local')
- goals - encode a stream of qubits
  - local stabilizers
  - hopefully still good performance

## classical convolutional codes:

linear system with finite memory, i.e.  
every output symbol depends linearly  
on the actual input symbol (or a group of inputs)  
and some previous inputs / the internal state  
of the encoder

Example: rate  $R = \frac{1}{2}$



linear feed-forward shift-register = polynomial multiplication  
feed-back = polynomial division

$$0 \dots 010 \dots 0 \xrightarrow{\oplus} 11 \dots 10 \dots 0 \quad \frac{1}{1+D} = \sum_{i \geq 0} D^i$$

input 1 produces an white sequence  
 $1+D$  produces the output 1

We use the powers of  $D$  as "time indicator", i.e.

$$\text{an input sequence } (a_0, a_1, \dots, a_\ell) \triangleq a_0 + a_1 D + a_2 D^2 + \dots + a_\ell D^\ell$$

## Main idea for error correction:

The linear shift-registers produce sequences which obey some linear "lineal" constraints

⇒ diagram of the internal state and the output symbols

memory  $m \Rightarrow 2^m$  internal states



- an edge between states  $s$  and  $s'$  labelled with the corresponding input/output
- every codesequence corresponds to paths in this trellis diagram
- error correction: finding the closest path given an output sequence

→ machine learning

## ⇒ Viterbi algorithm

basic idea  
compute for each state  $s$  the most likely sequence leading to this state, for memoryless channel,  
we can have a "local" update rule: consider only the last transition



for every state, we have computed the most likely path up to time  $t$

weight of a path

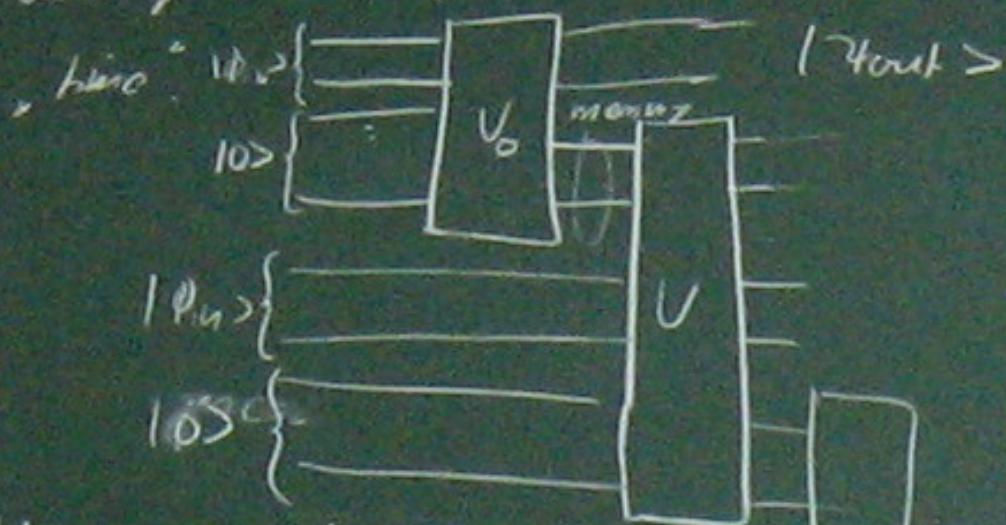
$$wgt(s_0 \rightarrow s) = wgt(s_0) + wgt('0')$$
$$wgt(s_1 \rightarrow s) = wgt(s_1) + wgt('1')$$

→ reflections

even with "small" memory, classical convolutional codes have a good performance (used in many comm. systems)

## Q.D.R. 2.0 Quantum Version

- a) use an operational analogue of shift-registers, e.g. use a fixed unitary (encoding circuit) and shift it in



With this approach, it is in general complicated to find an inverse encoding circuit and to correct errors.

encoderr

b) use an analogue of local constraints,  
i.e. stabilizers with finite support

Example:  $S_1 = \begin{array}{ccccccc} & \text{III} & \text{XXX} & \text{XZY} & \text{III} & \text{III} & \dots \\ & \text{III} & \text{ZZZ} & \text{ZYX} & \text{III} & \text{III} & \dots \end{array}$   
 $S_2 = \begin{array}{ccccccc} & \text{III} & \text{XXX} & \text{XZY} & \text{III} & \text{III} & \dots \\ & \text{III} & \text{ZZZ} & \text{ZYX} & \text{III} & \text{III} & \dots \end{array}$

shifted stabilizers  $S'_1 =$   
(by 3 qubits)  $S'_2 =$

$\Rightarrow$  define a semi-infinite stabilizer generated by  $S_1, S_2$   
and all shifts by multiples of three

$$G = \left[ \begin{array}{cc} \text{XXX} & \text{XZY} \\ \text{ZZZ} & \text{ZYX} \\ \text{XXX} & \text{XZY} \\ \text{ZZZ} & \text{ZYX} \end{array} \right] \Rightarrow \left[ \begin{array}{ccc|cc} 1+D & 1 & 1+D & 0 & D & D \\ 0 & D & D & 1+D & 1+D & 1 \end{array} \right] = \left[ \begin{array}{c|c} X(D) & Z(D) \end{array} \right]$$

max. degree of the polynomials + 1  
 $\geq$  number of qubits involved in

The code is the joint +1-eigenspace of a syndrome measurement  
all operators.

commutator conditions  
block codes:  $X^t Z^t - Z^t X^t = 0$   
convolutional codes  
 $X(D) Z(\frac{1}{D})^t$   
 $- Z(D) X(\frac{1}{D})^t = 0$

## 20.3 A bad example

$$G = \begin{bmatrix} Z & Z & Z \\ Z & Z & Z \\ Z & Z & Z \end{bmatrix} = (-0 \mid 1 + D)$$

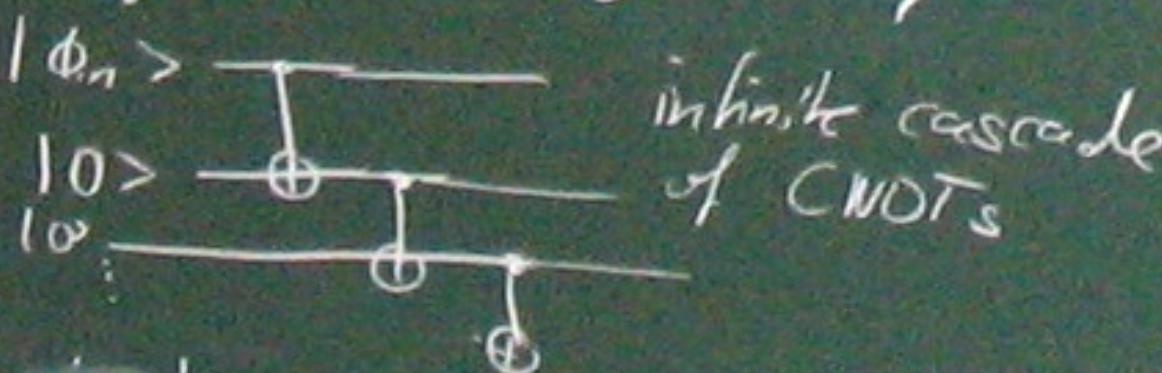
does not have  
a polynomial inverse

→ joint +1 eigenstates are

$$|\bar{0}\rangle = |00\rangle \quad >$$
$$|\bar{1}\rangle = |11\rangle \quad >$$

In particular  $|\bar{0}\rangle + |\bar{1}\rangle \in \mathcal{L}$

⇒ infinite, real "state  
encoding" would be given by



⇒ try to avoid these infinite depth  
circuits

## 20.4 Encoding Circuits

based on transforming the stabilizer

$$\left[ \begin{array}{c|c} X(D) & Z(D) \end{array} \right] \rightsquigarrow \left[ \begin{array}{c|cc} 0 & I & 0 \end{array} \right]$$

trivial block code

$\hat{=}$  convolutional code without  
memory

a) local operations  $H - \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$ ,  $P = \begin{pmatrix} 1 & 0 \\ 0 & i \end{pmatrix}$

$\Rightarrow$  operations on the columns of  $(X(D) | Z(D))$

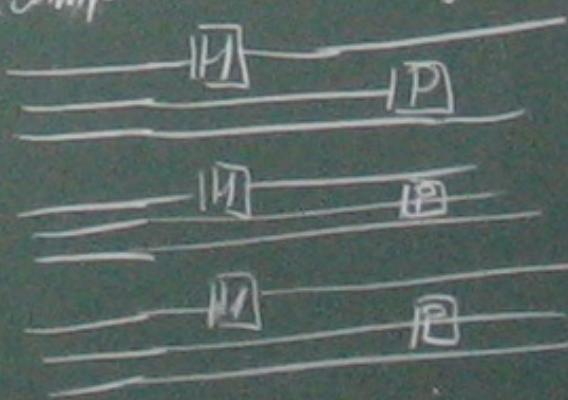
given by  $\tilde{H} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$   $\tilde{P} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$

$(X(D) | Z(D))$  corresponds to a semi-infinite

"block code" matrix

$\Rightarrow$  apply the gates  $H$  or  $P$  to all  
shifted blocks

example: block length 3



## References

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