Digital Communications, Advanced Course

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 ~ 250 patents in 4G-6G

 \sim 10000 Google Scholar citations

Current research interests: Large intelligent surfaces, Distributed MIMO processing

> Digital communications - Advanced course: Introduction - week 1

Wireless communication

Steady evolution from 1G to 5G

6G already discussed and planned for

More engineers/researchers employed now than ever

Enormous industry, large amounts of money involved

Enormously much more work to do, healthy industry

- Incorporation of AI
- Positioning/localization
- Base stations replaced with surfaces of electromagnetical surfaces
- Higher data rates
- THz
- etc

Digital communications - Advanced course: Introduction - week 1

- All systems, 1G 5G and the upcoming 6G system, based on the same principles
- Seen from a communication theory/mathematical perspective, they are almost the same system
- Important to understand said principles, no matter what part of the wireless industry one ends up in

Project work in this course

Option 1

- 2 students/group.
- A communication application/technical problem/problem area, relevant for the course, is investigated.
- The choice of project is mainly done by the project group.
- Articles and conference papers from IEEE's database "IEEE Xplore"

http://ieeexplore.ieee.org/Xplore/DynWel.jsp

is recommended to get additional technical information.

• Written report, oral presentation, and be opponent to another group.

Project work in this course

Option 2

- 2 students/group.
- Matlab based project related to MAP receivers and iterative decoding

Some examples of applications/systems studied in previous projects:

- Mobile telephony/broadband (GSM, EDGE, LTE, 4G, 5G,...)
- Internet
- Modem (e.g., ADSL)
- WLAN (Wireless Local Area Network)
- Digital TV
- MIMO
- Massive MIMO
- OFDM+MIMO
- GPS (Global Positioning System)
- Bluetooth
- mmWave

Overview of course

The course contains a number of important concepts, not all that closely related

- MAP receiver Important for iterative decoding, used in most modern systems
- Concept of signal space Used everywhere to transform continuous to discrete time
- Capacity Super important for communication deeper studied in information theory course
- Diversity Deeper studied in MIMO course
- Channels Deeper studied in channel modelling course
- OFDM Essential for WIFI, 4G and onwards
- Time varying channels Deeper str

Deeper studied in channel modelling course

Digital communications - Advanced course: Introduction - week 1 In reality, systems are using error correction coding.

In this course we investigate (mostly) uncoded systems as they are the basic building block for coded systems.

The course Coding Theory deals with coded systems.

MAP receiver

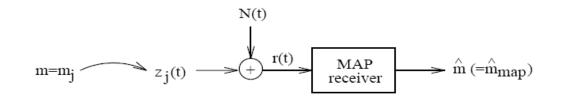
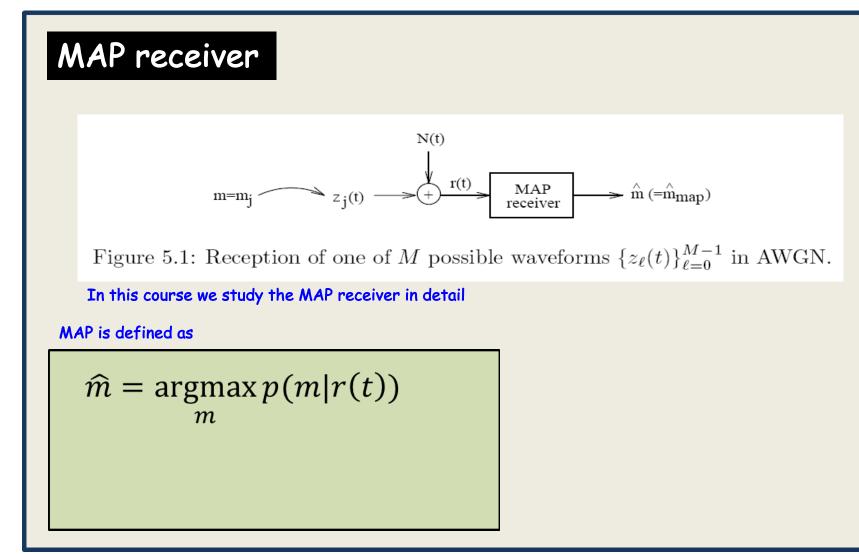
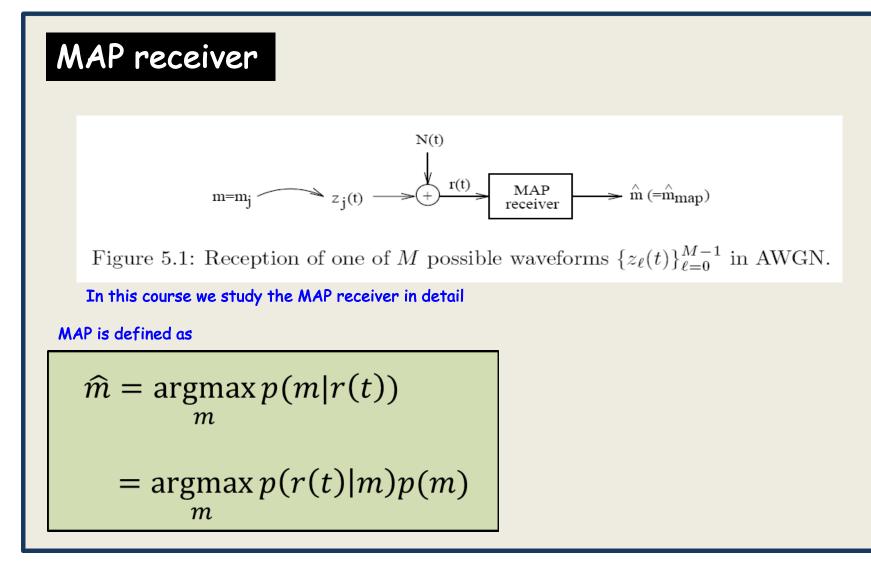
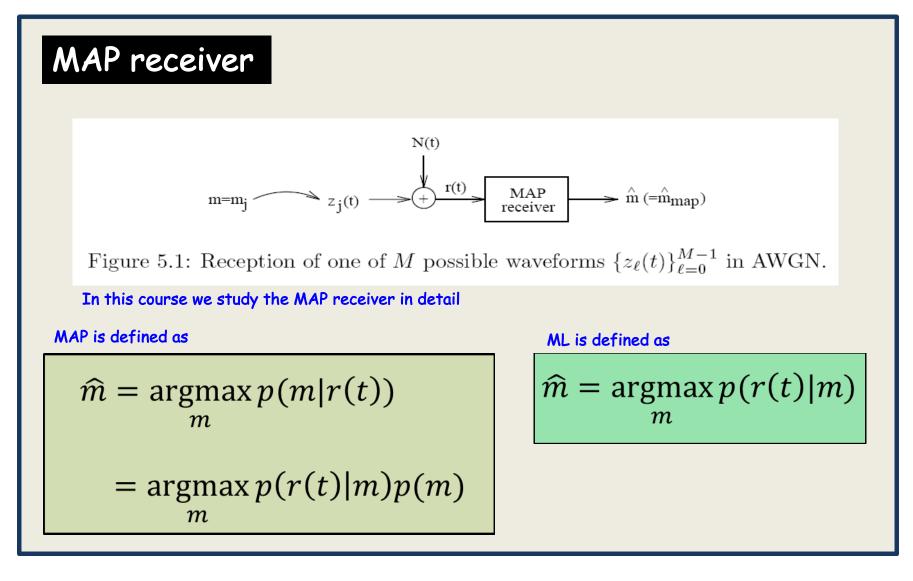


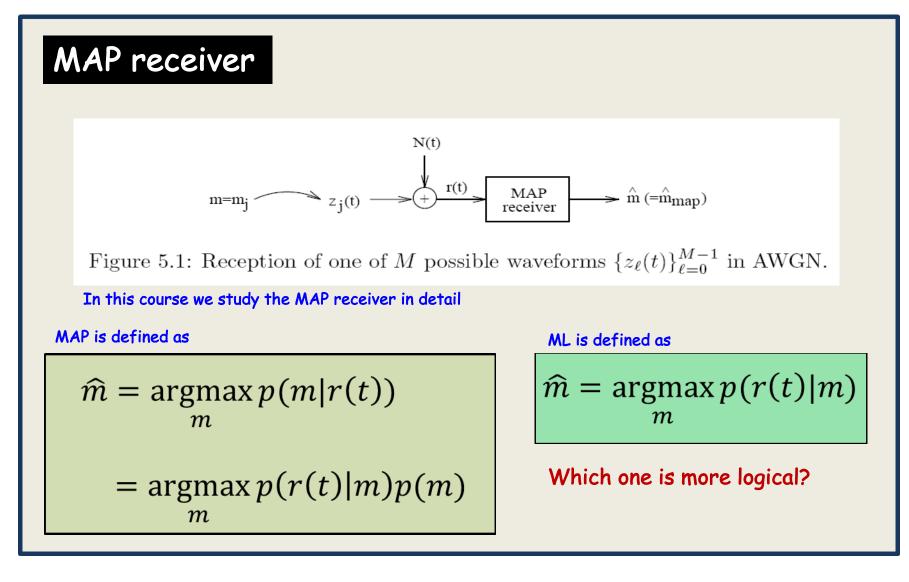
Figure 5.1: Reception of one of M possible waveforms $\{z_{\ell}(t)\}_{\ell=0}^{M-1}$ in AWGN.

In this course we study the MAP receiver in detail









MAP receiver

Suppose that on the next lecture, I will not show up

MAP is defined as

$$\widehat{m} = \operatorname*{argmax}_{m} p(m|r(t))$$
$$= \operatorname*{argmax}_{m} p(r(t)|m)p(m)$$

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MAP receiver

Suppose that on the next lecture, I will not show up Observation r(t) = "fredrik is not here"

Why am I not here?

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ML-rule p(r(t)|m₁) = 0.9 m₁ = "Fredrik is sick"

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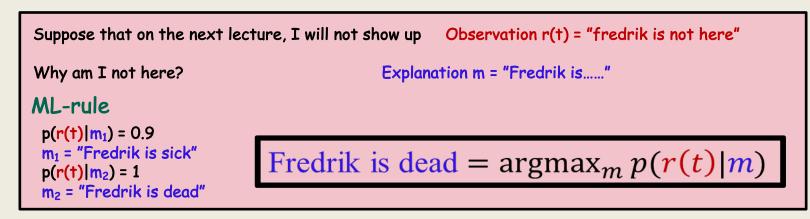
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MAP receiver

Fredrik is sick = $\operatorname{argmax}_m p(r(t)|m)$

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The ML rule is, in general, totally crazy

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- According to which rule should a court make their decisions?

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True court case: Sally clark case, England 1998

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Observation: Sally Clark, mother of two, had two babies that died in infancy

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True court case: Sally clark case, England 1998 MAP ? Observation: Sally Clark, mother of two, had two babies that died in infancy P(observation|natural causes) = 1/10000 according to expert in child deaths P(observation|murder) = 1 according to common sense P(mother is murderer) = 1/1000...000 P(mother is not murderer) = 0.9999...999

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MAP receiver

True court case: Sally clark case, England 1998 MAP: NOT GUILTY

Aftermath: Released in 2003, after some math professors took a look at the case.

Sally died from alhcolism somewhat later

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Hypotetical case: Lottery with 1000000 tickets

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ML: ??

MAP: ??

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ML: Jail

MAP: Prior probability of fraud must be evaluated

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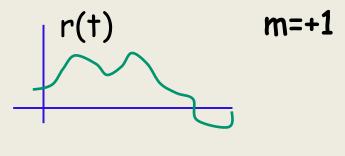
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To analyze a digital communications system, it is difficult to work with a continuous time signal

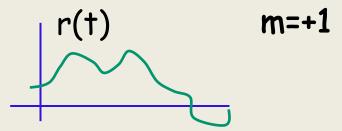
It is hard to evaluate a probability of the form p(r(t)|m)



MAP receiver

To analyze a digital communications system, it is difficult to work with a continuous time signal

It is hard to evaluate a probability of the form p(r(t)|m)



We are used to evaluate probabilites of the form p(r|m)

MAP receiver

To analyze a digital communications system, it is difficult to work with a continuous time signal

Concept of signal space:

- Transfer all continuous signals into discrete vectors
- Transformation should be such that no information is lost
- Transformation is done via a set of basis functions
- For two systems with identical signal spaces, all properties (Eb, BER, etc) are idetincal
- However, bandwidth properties are not. They depend on the basis functions
- Allows for a simpler description and analysis of the system.