

Modulation Classification of Satellite Communication Signals Using Cumulants and Neural Networks

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Automatic Modulation Classification



Objective

• Correctly predict the transmitted modulation scheme

Applications

- Automatic receiver reconfiguration
 - Reduce transmission overhead due to modulation coordination
- Interference Mitigation
 - Identify and respond to interferers uniquely
- Spectrum Management
 - Automate violation notification process



Requirements



Classify typical satellite communication signals

• *Ω* = {BPSK, QPSK, 8–PSK, 16–APSK, 32–APSK, 16–QAM, 64–QAM}

Evaluate performance with

- Various capture lengths
- AWGN, -5 to 20 dB
- Es/No approximation errors < 5 dB
- Phase and frequency offsets
- Nonlinear amplifier drive levels
- DVB-S2 pilots and headers

Assume

- Coarse carrier frequency estimation
- Symbol timing recovery
- Zero ISI, matched pulse shape filters



Classification Method



Cumulants

- Effective at differentiating modulation order
- Well documented in literature

Neural Networks

- Universal function approximator
- Showed increased accuracy over decision tree and SVM





Simulation Diagram

NA SA







Neural Network Architecture







What does the Neural Net see?







Vector Length Analysis



Feature vector generated from



For similar classification performance, classification based on $\{z[n]\}$ required ~15x more symbols

Frequency Offset



- Frequency offset imposes upper bound on y[n] sequence length
- z[n] converts fixed frequency offset into fixed phase offset
- Cumulant magnitudes are not impacted by constant phase offset



Es/No Approximation Error



- Neural net requires SNR estimation
- Imperfect estimation of SNR will degrade performance
- Most sensitive to error at low Es/No
- y[n] and z[n] exhibit similar responses to Es/No error
- Results provide accuracy requirements for SNR estimator



Nonlinear Amplifier



- Previous results in literature did not account for nonlinear amplification
- Amplifier simulated using Saleh model using coefficients from operational TWTA
- PSK only one ring, not impacted by amplifier
- Classification of higher order modulations experienced significant degradation at levels where a user could expect to operate
- Additional input features needed to train neural network over this dimension





DVB-S2 Pilots and Headers





- Previous research has not measured impact of pilots/headers on classifier performance
- DVB-S2 physical layer extends alphabet of received symbols, due to inclusion of headers/pilots
- Unable to classify 16 APSK using z[n] at 20 dB Es/No
- Classifier performance degradation due to DVB-S2 framing was < 5% in most cases



IQ constellations of 32 APSK with and without DVB-S2 physical layer



Next Steps and Conclusions



Next Steps

- Investigate additional features
- Implement a SNR approximation algorithm
- Classify modulation types in lab
- Add timing acquisition and carrier removal
- Classify live signals

Conclusions

- Created modulation classifier using cumulants and a neural network
- Evaluated performance over
 - Capture length
 - AWGN
 - Constant frequency and phase offset
- Extended previous work in field to include analysis over
 - SNR approximation error
 - Nonlinear amplifier distortion
 - DVB-S2 physical layer effects









Backup Slides





Classification by Modulation







Left: y[n] Right: z[n]



Cumulant Magnitudes







Left: y[n] Right: z[n]



DVB-S2 Pilots and Headers, Cont.



Probability of classifying modulation type with DVB-S2 headers (H) and pilots (P)

Es/No = 20 dB

z[n] signal type

