

3.1. Varadhan's Theorem

Ref: S.R.S. Varadhan, Comm. Pure & Appl. Math, 1966
↳ Abel Prize 2007

• Exponential expectation:

$$E[e^{nf(A_n)}] = \int e^{nf(a)} P(A_n=a) da \sim e^n$$

• Limit function:

$$\lambda[f] = \lim_{n \rightarrow \infty} \frac{1}{n} \ln E[e^{nf(A_n)}]$$

functional of f

• Assumptions:

1. A_n satisfies the LDP: $P(A_n=a) \asymp e^{-nI(a)}$

2. f bounded (can be relaxed)

Theorem:

$$\lambda[f] = \sup_a \{ f(a) - I(a) \}$$

"Proof":

$$E[e^{nf(A_n)}] = \int e^{nf(a)} P(A_n=a) da$$

$$\asymp \int e^{nf(a)} e^{-nI(a)} da$$

$$= \int e^{n\{f(a)-I(a)\}} da$$

$$\asymp e^{n \max\{f(a)-I(a)\}}$$

exponential
integral

Laplace principle
& approximation

$$\Rightarrow \lim_{n \rightarrow \infty} \frac{1}{n} \ln E[e^{nf(A_n)}] = \max\{f(a) - I(a)\}$$

□

$$E[e^{nkS_n}] = E[e^{k\sum_{i=1}^n X_i}] = E[e^{kX}]^n$$

3/

$$\Rightarrow \lambda(k) = \ln E[e^{kX}]$$

$$\text{But } E[e^{kX}] = e^{k\mu + \frac{\sigma^2}{2}k^2}$$

$$\Rightarrow \lambda(k) = k\mu + \frac{\sigma^2}{2}k^2$$

Question: Get I from λ ?

3.2. Gärtner-Ellis Theorem

Refs: J. Gärtner, Th. Prob. Appl. 1977

R.S. Ellis, Ann. Prob., 1984

• Scaled cumulant generating function:

$$\lambda(k) = \lim_{n \rightarrow \infty} \frac{1}{n} \ln E[e^{nk} A_n] \quad k \in \mathbb{R}.$$

• Assumptions:

◦ $\lambda(k) < \infty$

◦ $\lambda(k)$ is differentiable ~~forall~~ $k \in \mathbb{R}$ (Gärtner's version)

• Theorem:

① A_n satisfies the LPP: $P(A_n = a) \asymp e^{-nI(a)}$

② $I(a) = \sup_{k \in \mathbb{R}} \{ka - \lambda(k)\}$ Legendre-Fenchel transform

"Proof": cf HT2009, Appendix C

$$P(A_n = a) = \frac{1}{2\pi i} \int_B e^{-ka} M_{A_n}(k) dk$$

$$\asymp \int_B e^{-nka} e^{n\lambda(k)} dk$$

$$= \int_B e^{-n(ka - \lambda(k))} dk$$

$$\asymp e^{-n \min_k \{ka - \lambda(k)\}}$$
 along contour

$$= e^{-n \max_k \{ka - \lambda(k)\}}$$
 perpendicular to contour
for $k \in \mathbb{R}$

cf Appendix

$$\Rightarrow \lim_{n \rightarrow \infty} -\frac{1}{n} \ln P(A_n = a) = I(a) = \max_k \{ka - \lambda(k)\}$$

Rem:

- Why $\lambda(k)$ differentiable? See next week
- Are not necessarily iid sum - any RVs.

Example:

$$S_n = \frac{1}{n} \sum_{i=1}^n X_i \quad X_i \sim \mathcal{N}(\mu, \sigma^2) \text{ iid}$$

$$\begin{aligned} \lambda(k) &= \lim_{n \rightarrow \infty} \frac{1}{n} \ln E[e^{nkS_n}] = \ln E[e^{kX}] \\ &= k\mu + \frac{\sigma^2}{2} k^2 \end{aligned}$$

Differentiable $\forall k \in \mathbb{R}$.

$$\Rightarrow P(S_n = s) \asymp e^{-nI(s)}$$

$$I(s) = \sup_k \{ ks - \lambda(k) \}$$

$$\partial_k (ks - \lambda(k)) = 0 \Rightarrow s = \lambda'(k) = \mu + \frac{\sigma^2}{2} k$$

$$\Leftrightarrow k^* = \frac{s - \mu}{\frac{\sigma^2}{2}}$$

$$\Rightarrow I(s) = k^*s - \lambda(k^*)$$

$$= \left(\frac{s - \mu}{\frac{\sigma^2}{2}}\right)s - \left(\frac{s - \mu}{\frac{\sigma^2}{2}}\right)\mu + \frac{\sigma^2}{2} \frac{(s - \mu)^2}{\sigma^4}$$

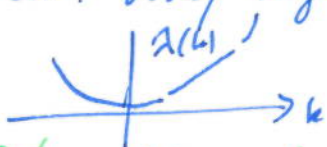
$$= \frac{(s - \mu)}{\sigma^2} (s - \mu) + \frac{(s - \mu)^2}{2\sigma^2}$$

$$= \frac{(s - \mu)^2}{2\sigma^2}$$

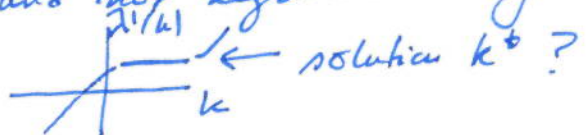
Inverse calculation compared to last calculation.

LF parabola = parabola

Rem: Why Legendre-Fenchel and not Legendre transform?



$$\lambda'(k) = s$$



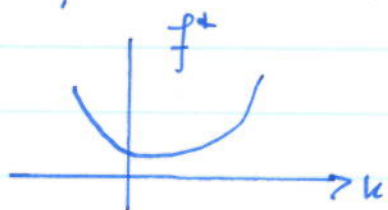
Other examples next week.

3.3. Legendre - Fenchel transform

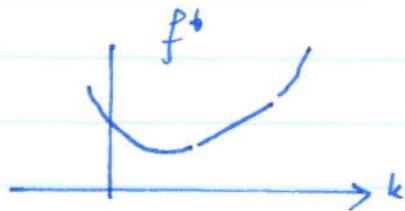
$$f: \mathbb{R} \rightarrow \mathbb{R}$$

$$\text{Def: } f^*(k) = \sup_x \{ kx - f(x) \} \quad k \in \mathbb{R}$$

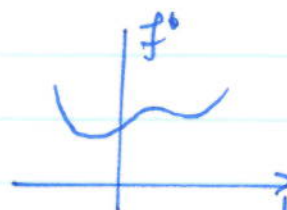
- $f^*(k)$ is a convex function



convex



convex, non-strictly



non-convex

- Inverse LF transform: If $f(x)$ is convex, then

$$f(x) = \sup_k \{ kx - f^*(k) \}$$

- LF is self-inverse (dual or involutive) for convex functions

- Legendre transform: If $f(x)$ is differentiable and strictly convex, then

$$f^*(k) = kx_k^* - f(x_k^*) \quad f'(x) = k$$

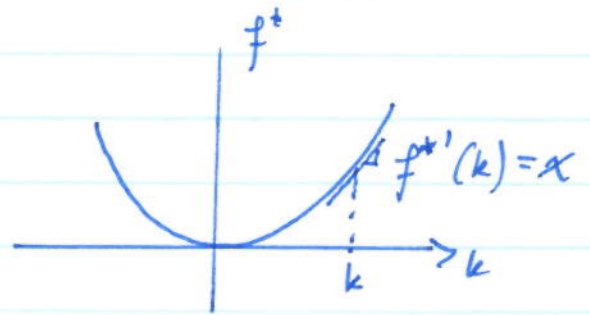
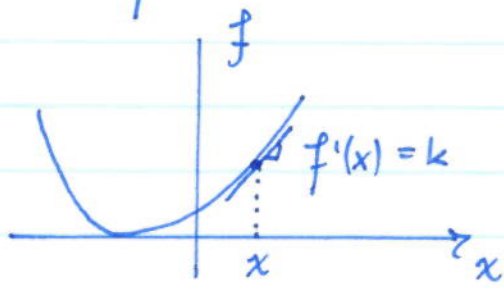
$$\text{Proof: } f \text{ diff: } \partial_x (kx - f(x)) = 0 \Rightarrow f'(x) = k$$

$$f \text{ strictly convex: } f'(x) = k \text{ has unique root } x_k^* \quad \forall k \in \mathbb{R}$$

$$\Rightarrow f^*(k) = kx_k^* - f(x_k^*). \quad \square$$

Rem: LF more general because can be applied to nondifferentiable and/or nonconvex functions.

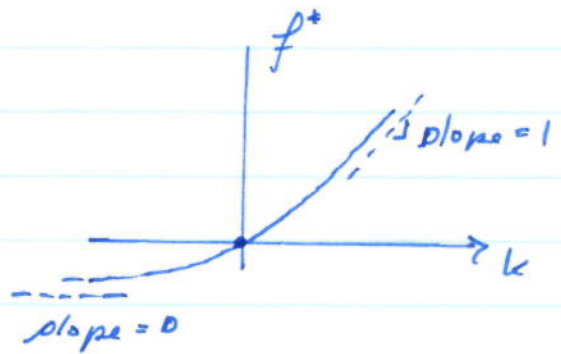
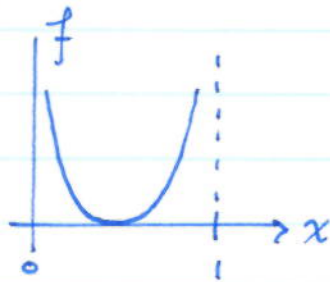
• Duality:



$$f' \leftrightarrow k$$

$$f^{*'} \leftrightarrow x$$

Example:



Example:

$$f(x) = \frac{(x-a)^2}{2b}$$

$$f^*(k) = ak + \frac{b}{2}k^2$$

The only fix point.

3.4. Properties of $\lambda(k)$ and $I(s)$

• $\lambda(0) = k$

Proof: $\lambda(k) = \lim_{n \rightarrow \infty} \frac{1}{n} \ln E[e^{nk S_n}]$

$$\lambda(0) = \lim_{n \rightarrow \infty} \frac{1}{n} \ln E[\underbrace{1}_{=1}]$$

iid case: $\lambda(k) = \ln E[e^{kX}]$

$$\lambda(0) = 0$$

• $\lambda'(0) = E[X]$ iid case

• $\lambda''(0) = \text{var}(X)$ iid case

• $\lambda'(k) = \lim_{n \rightarrow \infty} E[S_n]$ in general

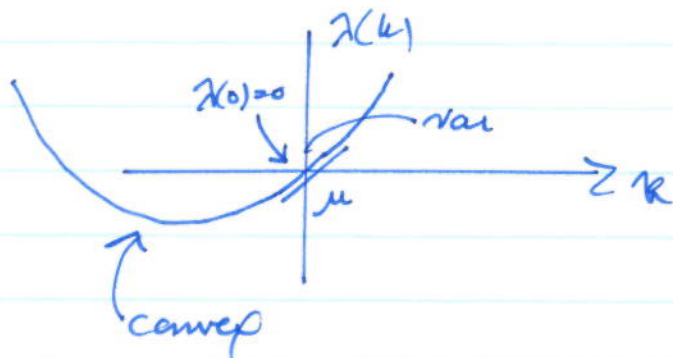
• $\lambda(k)$ is convex

Exercise

• Rate functions obtained from GE are convex.

Rem: $I(s)$ may be nonconvex

Example next week



None from exercises